

Analysis of Washington Nutrient and Biological Data (Periphyton) for the Nutrient Scientific Technical Exchange Partnership Support (N-STEPS)

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1. INTRODUCTION

This report summarizes the results of the workplan developed to analyze existing water chemistry and periphyton data from the State of Washington. The main goal of this research was to evaluate how well periphyton assemblages could be used as a screening tool to identify excess nutrient conditions in Washington streams. In addition, because natural variability due to biogeographic gradients in nutrient or nutrient-response relationships introduces controllable error, the effects of a priori classes (e.g., ecoregions) and available natural gradient factors were explored for potential division of Washington sites into separate classes. If classes are identified, classification will reduce variability in distributional statistics and improve fit of stressor-response models. It builds off of an effort to look at similar responsiveness of periphyton metrics to nutrient gradients in Oregon, incorporating additional data from the State of Washington, Department of Ecology.

The proposed outcome of the analysis plan was to develop a brief report (this document) that provided the following:

1. A conceptual model for stressor-response relationships for the stream types (subclasses) that WDOE identifies.
2. A GIS map of relevant locations with applicable data throughout the western ecoregions.
3. Site classes and co-varying environmental variables used to reduce natural variability in the nutrient data developed from the broader (multi-state) dataset.
4. Analysis of the above biological samples (periphyton identification and quantification), including:
 - Endpoints determined from frequency distribution analysis for each nutrient type and form, including cumulative distribution functions and the range of standard distributional statistics (mean, variance, standard deviation, standard errors, coefficient of variation, median, and quartiles);
 - Endpoints from stressor response analysis including visual plots of interest with loess curve fits, interpolated linear regression stressor values for response of interest, and thresholds determined for non-linear relationships using change-point analysis. Uncertainty statistics will be provided for each estimate;
 - A qualitative discussion of data gaps and the basis for any recommendations. Gaps discussion will focus on ranges of gradients and physical locations, but will also include analysis gaps that could be addressed with future data, modeling, or field collection effort;
5. A database containing relevant data that is gleaned from reported studies/assessments.
6. All R code, annotated with detail, for analyses will also be provided (not included in this report, but provided separately)

Item 1 was provided as a series of powerpoint presentations on the conceptual model. This report represents provision of items 2-4, and items 5 and 6 will be provided or made available along with this report as requested.

The report briefly discusses methods, results, and provides a summary of the data analysis. Example plots and tables are shown in the main text, additional data and analysis output are provided in the appendices.

2. METHODS

Study area

The initial study area included the state of Washington and surrounding sampling stations that are located within Omernik level III ecoregions (USEPA, 2013) that are also contained within the state of Washington (Figure 1). After preliminary classification analysis, however, in which latitude was shown to have an influence on diatom metric vs. nutrient concentration model residuals and concern about extensive distance creating non-relevant effects on community structure, California sites (latitude < 42°N) were removed from subsequent analyses and analyses were repeated without those sites.

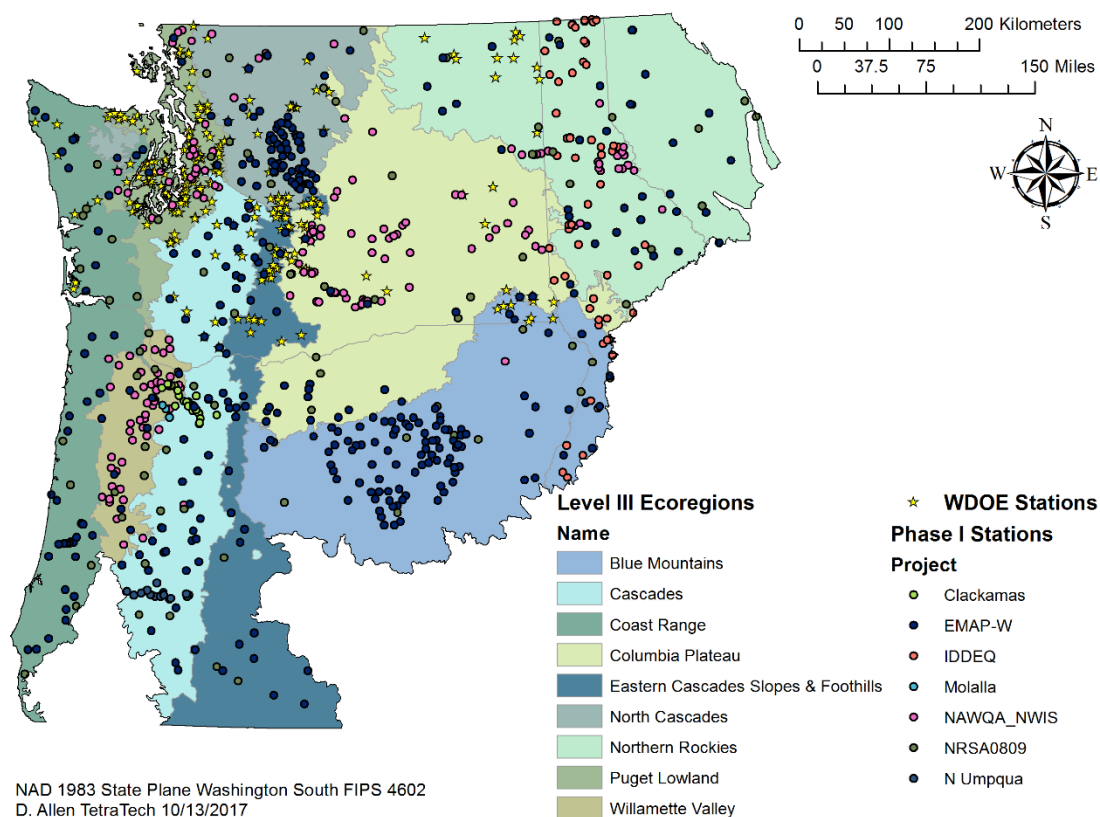


Figure 1 - Sampling stations (N=856) included in the current analysis in reference to Washington ecoregions.

Data preparation

The stream chemistry, nutrient, and periphyton data were obtained from two sources: 1) Washington Department of Ecology (WDOE) and 2) a previous NSTEPS project developed for the state of Oregon which contained data from several sources (Table 1). The WDOE data were merged with the previous NSTEPS project data. Water quality parameters used in this analysis were limited to those provided by these sources (Table 2). Because EPA and USGS projects used different nomenclature, chemical parameters collected in those projects were aggregated (Appendix 1). Some values derived from older analytical techniques were excluded as the values are not comparable to newer methods. Samples collected earlier than 2000 were also excluded to better match the WDOE dataset, whose samples were collected from 2010 - 2015. Samples whose shipping or holding times exceeded recommended ranges were included. Because reported values were frequently below detection limits, and detection limits varied by project and within parameters, all values were included as recorded in the original datasets. For the WDOE data, the values were used as provided without modification.

All chemical parameters except pH, dissolved oxygen (DO), and temperature were log transformed. Averages therefore often represent geometric means. The final dataset prepared by the methods described above included 4668 samples from 856 distinct sampling sites.

Table 1 – Source of data used in the analysis.

Agency	Project	Data Source
Washington Department of Ecology	Multiple	WDOE, Chad Larson
US EPA	Western EMAP (Environmental Monitoring and Assessment Program)	http://www.epa.gov/wed/pages/models/EMAP_West_Data.htm
	NRSA0809 (National Rivers and Streams Assessment)	http://water.epa.gov/type/rsl/monitoring/riverssurvey/index.cfm
USGS Special Projects	Clackamas	USGS, Kurt Carpenter
	Molalla	USGS, Kurt Carpenter
	N. Umpqua	USGS, Kurt Carpenter
USGS	NAWQA (National Water Quality Assessment)	USGS, Kurt Carpenter
Idaho Department of Environmental Quality		Tetra Tech, Ben Jessup

For the periphyton data collected by non-WDOE entities, two or three samples had frequently been collected on each visit to a given site. To avoid duplication, only one sample per site visit was chosen to include in the analysis. For those samples for which the sample type was

specified, types ID and ARTH were included. For samples collected at the same site on the same day, samples labelled “primary” or “A” were included and samples labelled “duplicate” or “B” were excluded. Diatom counts including less than 500 or more than 700 total valves counted were excluded. Several samples included duplicate species counts; those counts were summed.

Additional diatom sample preparation unique to the WDOE samples included: adding a code for diatom only samples, removing all NA taxa from consideration, fixing SampleID - QUAR77-2015-0715-10:11 so as to be included, and merging taxa that had two separate enumerations within a sample for the same taxon.

After all taxonomic data from WDOE were merged with the previous data, a taxonomic reconciliation effort was undertaken. Because samples were obtained from several labs, nomenclature varied or ambiguous taxonomies were often used (e.g., resolution of some individuals to genus and others to species within a genus, etc.). Therefore, these differences had to be resolved and the taxonomy converted to the nomenclature tied to the diatom attribute and trait information used to calculate metrics.

Periphyton data were collected into one table. In general, reported taxa containing “cf.,” “aff.,” or “sp.” with or without a number or question mark, were considered ambiguous identifications and were assigned the lowest taxonomic identification possible (usually genus). Reported taxa that were unambiguously identified were compared to the Academy of Natural Sciences [ANS; NADED (North American Diatom Ecological Database) fields] taxa lists and to taxonomic names previously reviewed by Tetra Tech. If a taxon name appeared on the ANS list, the taxon was assigned that name. If not, but it did appear on the Tetra Tech list, it was assigned that name. If it appeared on neither list, it was compared to the California Academy of Sciences (CAAS) database, and if a match was found, was assigned that taxonomic name. Finally, remaining taxa were searched for on ITIS (www.itis.gov) and AlgaeBase (www.algaebase.org) to identify possible synonyms or basionyms that might be present on one of the above lists. If no matches were found, the taxon was resolved to genus if possible or excluded if not. After taxon reconciliation above, some taxa were listed twice, for example, two separate taxa resolved up the same genus. Those counts were combined, and then total counts and relative abundances were recalculated for all the samples. In the end, 46 taxa remained unresolved and were removed, these were rare taxa however and represented 0.3% of individual valves and affected only 0.5% of the samples.

From the diatom data, 30 metrics were calculated using an Access database developed for a prior project (Appendix 2). These primarily consisted of two main metric types: composition (“percent individuals”) and weighted average based. The composition metrics are calculated as the percent of individual diatoms in the sample reflecting a particular trait, for example benthic or sestonic for the BEN_SES metric. The weighted average based metric are metrics that are calculated using weighted average optima of taxa to a specific nutrient or stressor (e.g., total phosphorus, TP) such as the wa_OptCat_L1Ptl metric (weighted average TP from diatom optima) or general environmental gradient (e.g., watershed development, MVI) such as the wa_OptCat_NutMMI (weighted average multivariate nutrient optima). Weighted average metrics were also calculated for some specific trait categories such as MOISTURE, pH, POLL_TOL where the result is the weighted average category value based on diatom abundances and their specific trait values for dry tolerance, pH, and pollution tolerance, respectively.

To match periphyton data to water chemistry samples, a window of 28 days before and 14 days after was used to choose candidate chemistry samples. From that window, the sample closest in time to the periphyton sample collection date was manually selected. For samples without a collection date (USGS special projects), samples were matched manually by year. The dataset described above containing the periphyton assemblage metrics and paired chemistry data contained 880 samples representing 674 distinct sampling sites (Figure 2-not made yet, the bubble plot you wanted).

Lastly, for classification analysis, physical, geographic and landscape-level variables were collected. Precipitation was extracted from the PRISM Climate Group 30 year (1981-2010) normal annual precipitation raster (2012). For landscape level characteristics (i.e. forest cover), WDOE provided StreamCat data, which provides summary statistics based on the watershed area of each NHD reach on which a sampling station is located. For the non-WDOE data, watersheds had been previously delineated to calculate landcover characteristics based on the watershed upstream of a given sampling sites. For example, for sampling stations located near the downstream node of an NHD reach, the delineated watershed would provide nearly identical results to the data provided by StreamCat for that reach. For a sampling site further upstream from the outlet, the delineated watershed would not include contributing watershed area to the reach downstream of the site in the calculated land cover data.

Data description, visualization and frequency distribution

Descriptive statistics were calculated for each parameter. Boxplots and cumulative distribution functions were created for all parameters for all the study sites divided by reference designation, and for total phosphorus (TP) and total nitrogen (TN) for study sites divided by ecoregion. Reference site designations were used as attributed by WDOE, EMAP-W and NRSA in their datasets. However, because some projects from the previous NSTEMPS database did not include a reference designation, some of the “Other” category streams may be from relatively undisturbed watersheds. Samples from the Omernik Level III Ecoregion 2 (Puget Lowland) were plotted separately as their reference sites had higher TN concentrations than reference sites from other ecoregions.

Stressor-Response

Spearman correlation analysis was used to explore relationships between biological response periphyton metrics and nutrient concentrations. Loess (locally weighted scatterplot smoothing) fits were used to explore the shape of relationships. Then, linear models were developed to determine stressor-response relationships. Where statistically significant ($p < 0.05$) nutrient-periphyton relationships existed, the nutrient concentrations associated with response targets calculated from reference sites (10th and 25th percentiles for linear models with negative slopes and 75th and 90th percentiles for linear models with positive slopes) were interpolated. Significant correlations ($p > 0.05$), were selected from among the periphyton metric response models and used for guiding subsequent classification analyses.

Following classification analysis, stressor response analysis was repeated for the candidate classes.

Classification

Natural variability due to biogeographic gradients in nutrient or nutrient-response relationships introduces controllable error. Therefore, the effect of a priori classes (e.g., ecoregions) and available natural gradient factors were explored for potential classifications to reduce variability in distributional statistics or stressor-response models. Three general techniques were used to assess the effect of ecoregions and natural gradients on model relationships: 1) visual analysis of model residuals and 2) model-based recursive partitioning, and 3) Nash-Sutcliffe efficiency calculation. Natural gradients included elevation, precipitation, longitude, latitude and percent forest cover.

To examine model residuals in relation to environmental gradients, the periphyton metrics with the highest Spearman correlations to nutrient concentrations were chosen to develop test models. Those models were run with all the data, and model residuals were examined as a function of ecoregion, longitude, latitude, elevation, and precipitation.

Model-based recursive partitioning was used to determine if and how biogeographic gradients and ecoregion classes affect the linear regression model between nutrient concentrations and periphyton metrics (Alexander and Grimshaw, 1996). Model-based recursive partitioning splits samples into subsets, based on a specified continuous or categorical covariate, in which the relationship between the stressor and response variables has minimal summed deviance across the individual models for all subsets of the data as compared to the deviance of the original model built from unsplit data (Zeileis et al 2008). To implement model-based partitioning, the partykit package (Hothorn and Zeileis 2014) in R was used (R Core Team 2014).

After determining possible classes, Nash-Sutcliffe efficiency (NSE) was calculated to quantify how well diatom metrics - nutrient concentration models built from one class described relationships from sites of a different class. The NSE is a normalized statistic that determines the relative magnitude of a test group residual variance compared to the measured data variance with a “1” being a perfect match of modelled to observed data, and values less than one indicating that the data mean is a better predictor than the modelled values (Nash and Sutcliffe 1970). It is calculated by:

$$NSE = 1 - (\text{sum}((obs - sim)^2) / \text{sum}((obs - \text{mean}(obs))^2)$$

NSE was calculated for diatom metric vs. nutrient concentration models for ecoregions 3, 10, and 11 combined, to which samples from remaining ecoregions were compared and NSE of that test data compared to the ecoregion 3, 10, and 11 model. The hydroGOF package was used to calculate NSE (Zambrano-Bigiarini 2014).

3. RESULTS -

Data description, visualization and frequency distribution

Exploratory data visualization revealed that reference sites in the Puget Lowland ecoregion had higher TN values than other ecoregions, although values across all sites did not vary as much

(Figure 2). WDOE confirmed that due to widespread urbanization in the Puget Lowland ecoregion, it was necessary to relax standards to some degree to approximate reference conditions. Therefore, in summarizing nutrient concentrations by reference and non-reference populations, ecoregion 2 reference sites were evaluated separately. Ecoregions 1, 10 and 11 tended to have higher TN concentrations in reference sites, whereas TP values in reference sites were higher at ecoregion 9, 10, and 11 sites (Figure 2 and Figure 3). Ecoregion specific cumulative frequency distributions across all sites (including non-reference) were otherwise characterized by higher nutrients in ecoregions 3 and 10 (Willamette Valley and Columbia Plateau).

Descriptive statistics of grab samples for each parameter from all sites, reference sites and reference sites in ecoregion 2 sites are summarized in Table 2 and graphically in Figure 4 to Figure 7. As expected, nutrient concentrations are lower in reference sites (not including ecoregion 2). Ecoregion 2 reference sites had higher mean chloride (Cl), total suspended solids (TSS), specific conductivity, turbidity, and benthic chlorophyll *a* biomass than sites not designated reference.

Stressor response

Benthic chlorophyll *a* had no apparent relationship with nutrient concentration (Figure 8). Diatom metrics developed based on diatom nutrient optima, however, responded as expected to nutrient concentrations, increasing with TN and TP respectively (Figure 9 and 10, and see Appendix 3). For the periphyton metrics, reference site quartiles and deciles were calculated as response targets and TN and TP values associated with these targets were interpolated from the simple linear regression models using the TN and TP concentrations at which response target values intersected the mean regression line (Table 3 and 4).

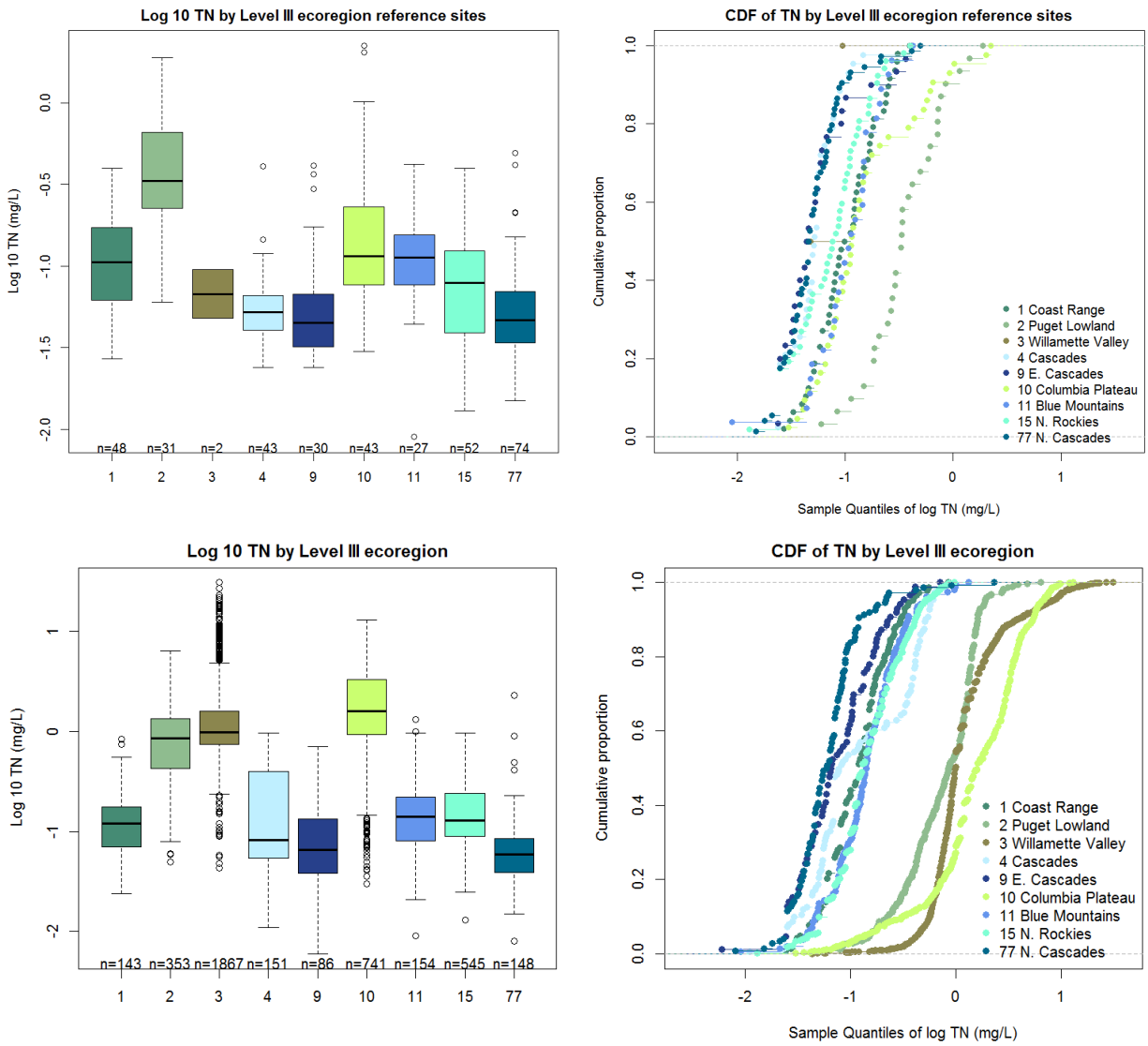


Figure 2 - Boxplots (left) and cumulative distribution functions (right) of log₁₀ transformed TN from reference sampling stations (top) and all sites (bottom) divided by Level III ecoregion.

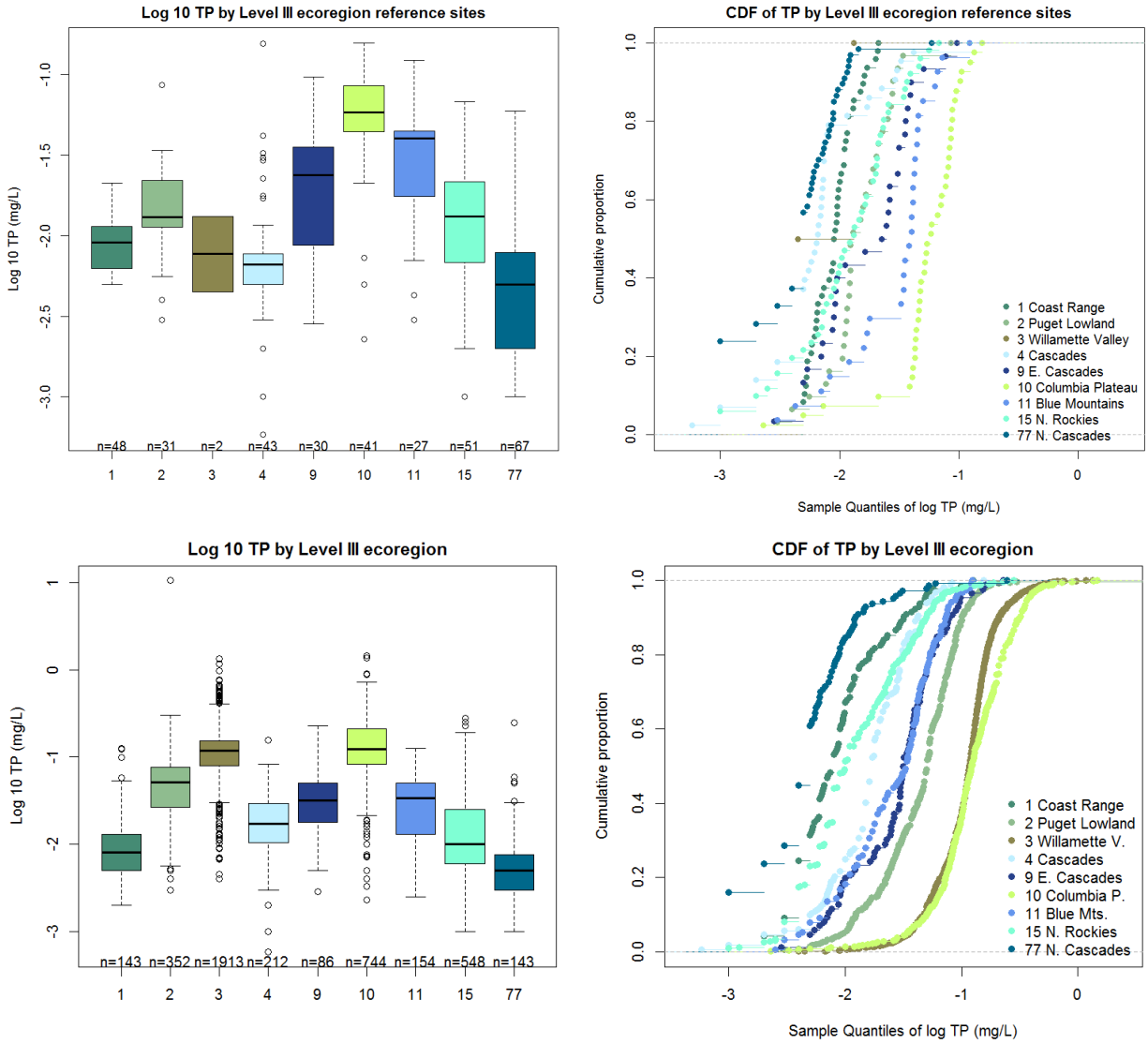


Figure 3 - Boxplots (left) and cumulative distribution functions (right) of \log_{10} transformed TP from reference sampling stations (top) and all sites (bottom) divided by Level III ecoregion.

Table 2 - Descriptive statistics of grab sample nutrient and benthic chlorophyll a (Chl a) data for all sites, reference sites and reference sites from ecoregion 2 from all months. Parameter abbreviations described in Appendix 1.

Analyte	Group	N	Mean	10th	25th	50th	75th	90th
Cl (mg/L)	All	859	1.085	0.241	0.44	0.918	2.989	6.306
	Reference	313	0.693	0.177	0.308	0.53	1.29	4.974
	Reference, ecoregion2	31	2.836	1.1	1.47	1.94	6.405	13.9
TN (mg/L)	All	4188	0.605	0.076	0.221	0.79	1.4	3.2
	Reference	319	0.075	0.025	0.044	0.068	0.12	0.228
	Reference, ecoregion2	31	0.358	0.149	0.226	0.333	0.663	0.854
TP (mg/L)	All	4295	0.054	0.01	0.03	0.08	0.13	0.2
	Reference	309	0.011	0	0.01	0.01	0.03	0.05
	Reference, ecoregion2	31	0.015	0.01	0.01	0.01	0.02	0.03
TSS (mg/L)	All	841	1.952	0.5	1	2	4	8
	Reference	305	1.576	0.5	1	1	3	6
	Reference, ecoregion2	31	2.411	1	1	2	5.5	8
Conductivity (µS/cm)	All	931	88.54	34	54.78	83.9	153	256
	Reference	310	67.599	23.25	39.81	69.34	113.27	182.22
	Reference, ecoregion2	25	98.66	57.6	79.6	89.5	121.15	186.9
DO (mg/L)	All	434	9.616	8.35	9.14	9.8	10.37	10.85
	Reference	217	9.899	8.94	9.4	9.92	10.5	11.06
	Reference, ecoregion2	26	9.9427	9.46	9.8	9.88	10.38	10.61
pH	All	891	7.679	7.1	7.4	7.67	7.99	8.28
	Reference	312	7.608	7.04	7.39	7.63	7.91	8.14
	Reference, ecoregion2	29	7.464	7.1	7.35	7.54	7.69	7.9
Turbidity (NTU)	All	731	0.724	0.1	0.3	0.7	1.8	3.3
	Reference	259	0.57	0.1	0.2	0.5	1.4	2.6
	Reference, ecoregion2	24	0.933	0.3	0.6	1.2	1.9	2.4
Chl a (mg/m ²)	All	626	15.325	3	6	16	41.1	79.8
	Reference	221	12.915	3.8	6.2	12.7	25.2	57.7
	Reference, ecoregion2	28	28.180	9.2	16.7	27.3	46.1	113.2

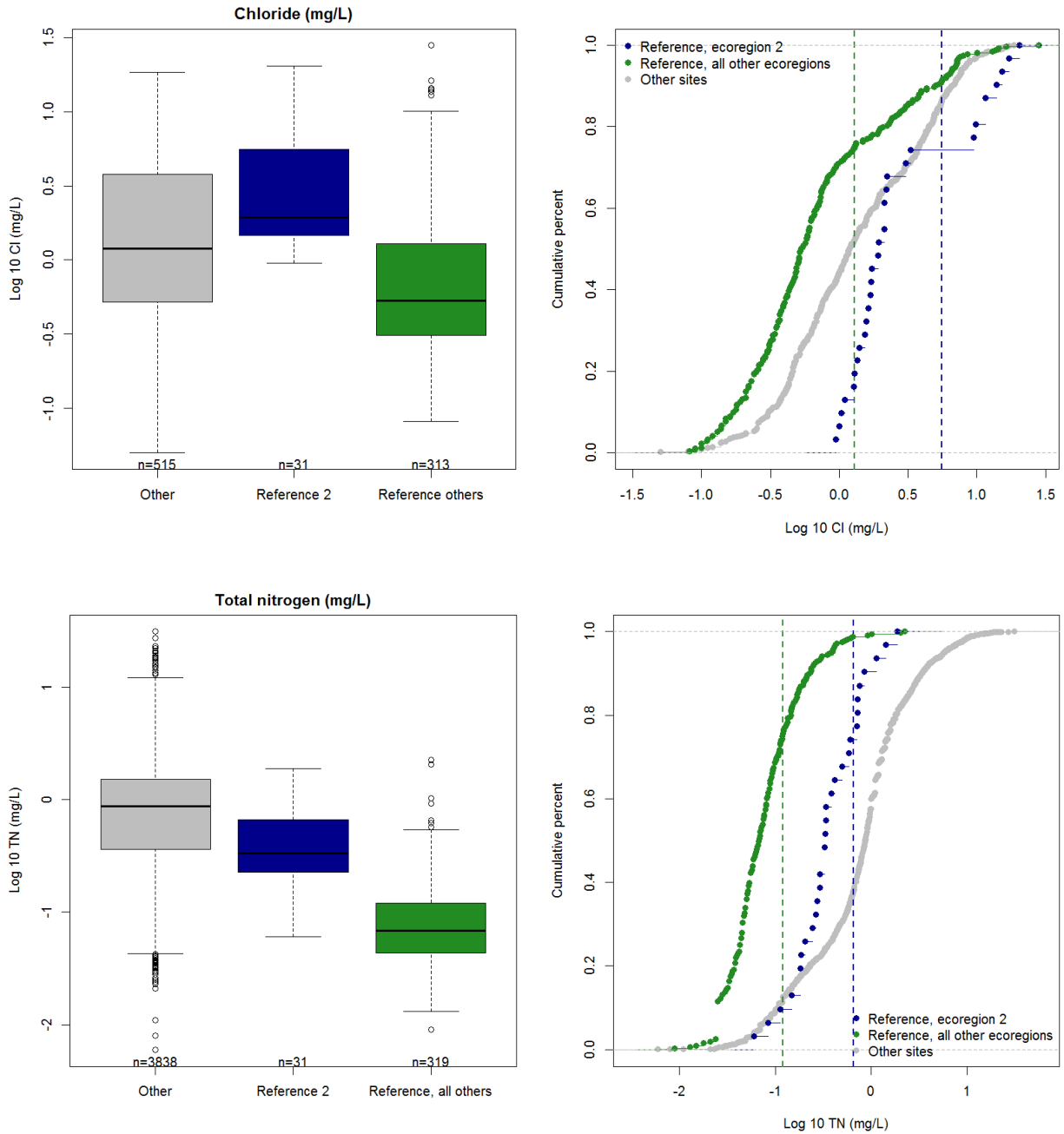


Figure 4 - Distributions of grab sample chloride (top) and TN (bottom) in streams not specifically designated as reference (“Other”, grey symbols) and in reference streams, presented separately for Omernik Level III ecoregion 2 (blue symbols) and all other ecoregions (green symbols). The green dashed vertical line is the reference site 75th percentile for non-ecoregion 2 ecoregions. The dark blue dashed vertical lines is the 75th percentiles of ecoregion 2. Sample size for these figures are provided in the boxplots and also in Table 2.

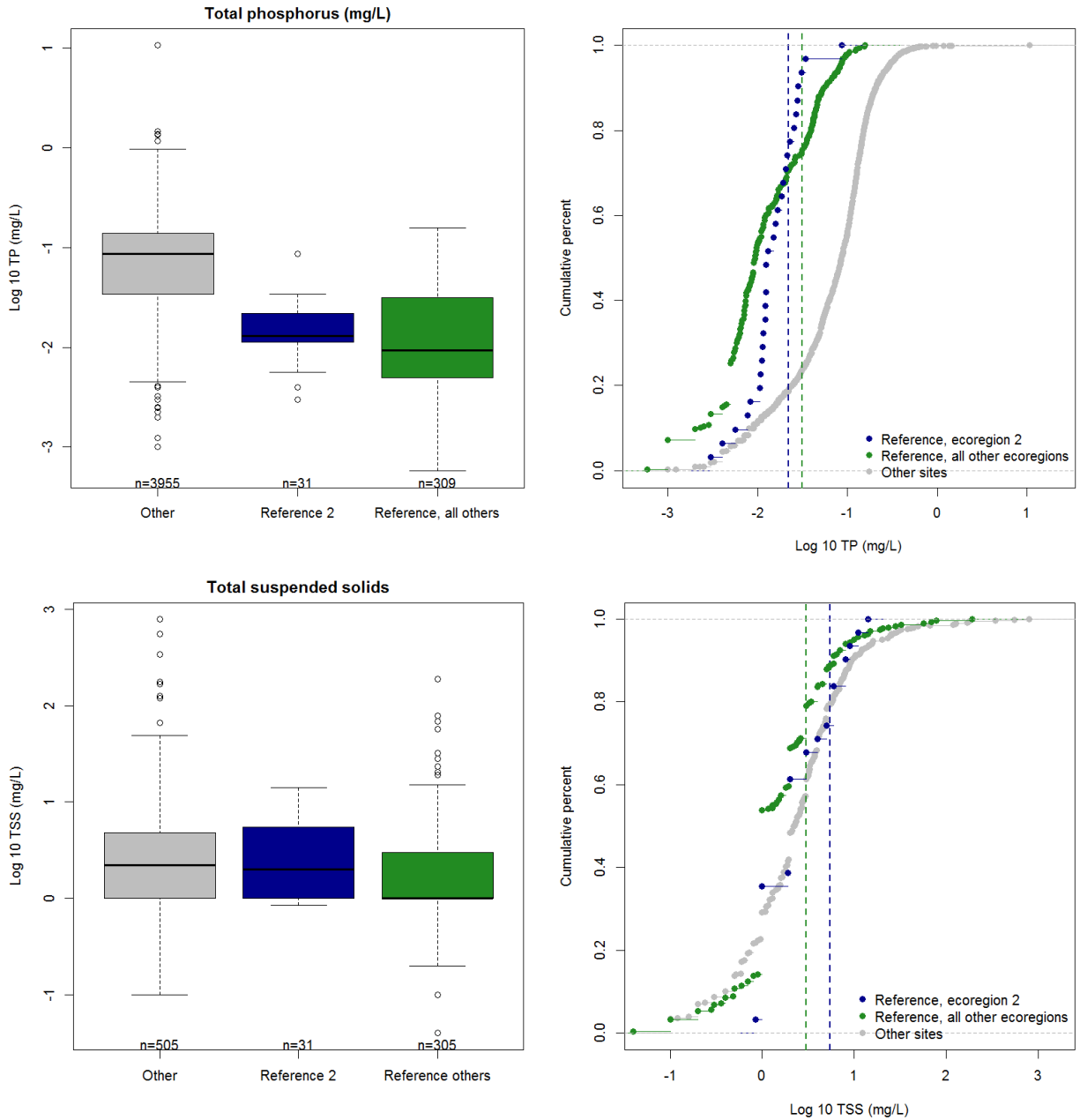


Figure 5 - Distributions of grab sample TP (top) and TSS (bottom) in streams not specifically designated as reference (“Other”, grey symbols) and in reference streams, presented separately for Omernik Level III ecoregion 2 (blue symbols) and all other ecoregions (green symbols). The green dashed vertical line is the reference site 75th percentile for non-ecoregion 2 ecoregions. The dark blue dashed vertical lines is the 75th percentiles of ecoregion 2. Sample size for these figures are provided in the boxplots and also in Table 2.

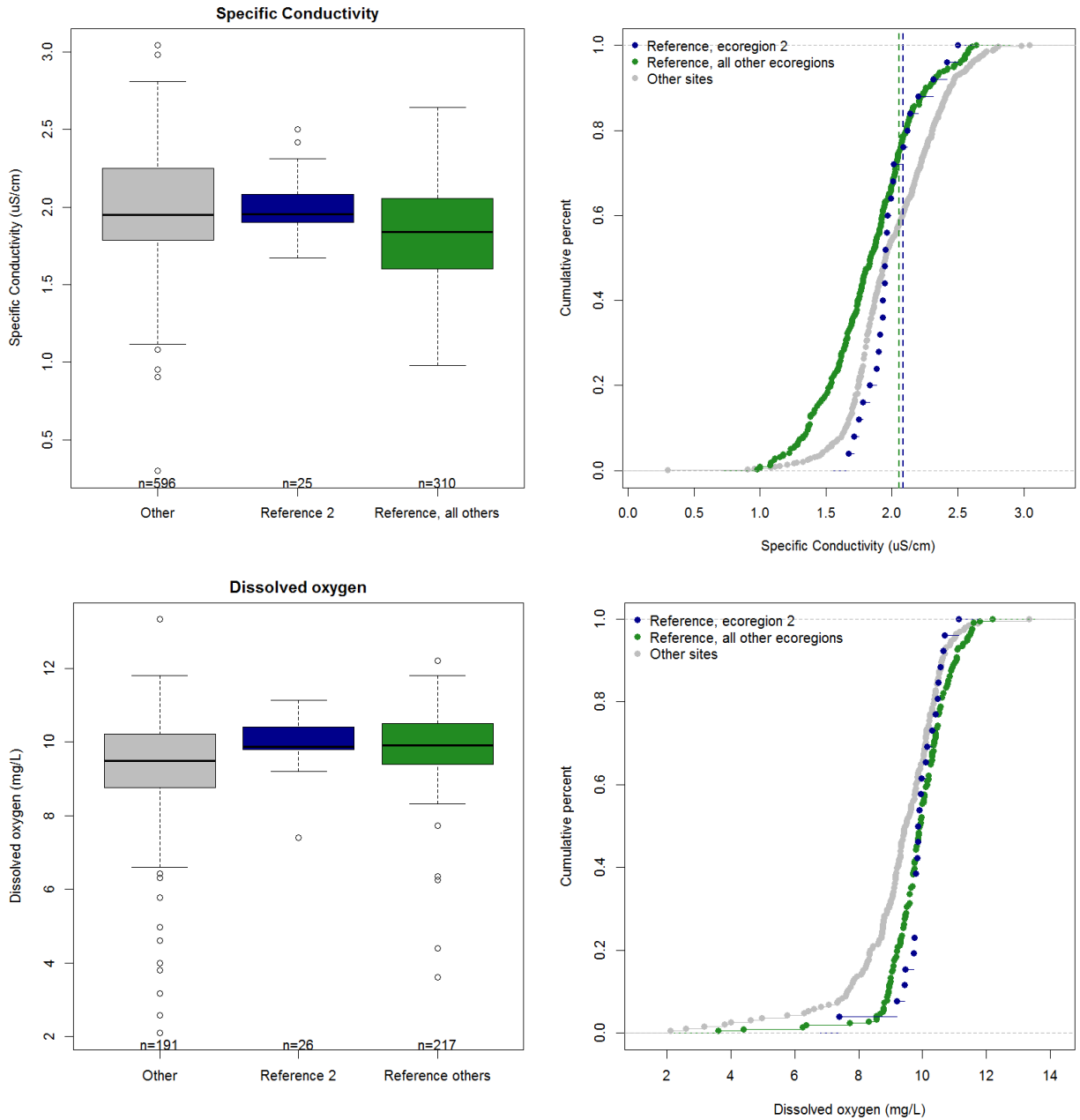


Figure 6 - Distributions of grab sample specific conductivity (top) and dissolved oxygen (bottom) in streams not specifically designated as reference (“Other”, grey symbols) and in reference streams, presented separately for Omernik Level III ecoregion 2 (blue symbols) and all other ecoregions (green symbols). The green dashed vertical line is the reference site 75th percentile for non-ecoregion 2 ecoregions. The dark blue dashed vertical lines is the 75th percentiles of ecoregion 2. Sample size for these figures are provided in the boxplots and also in Table 2.

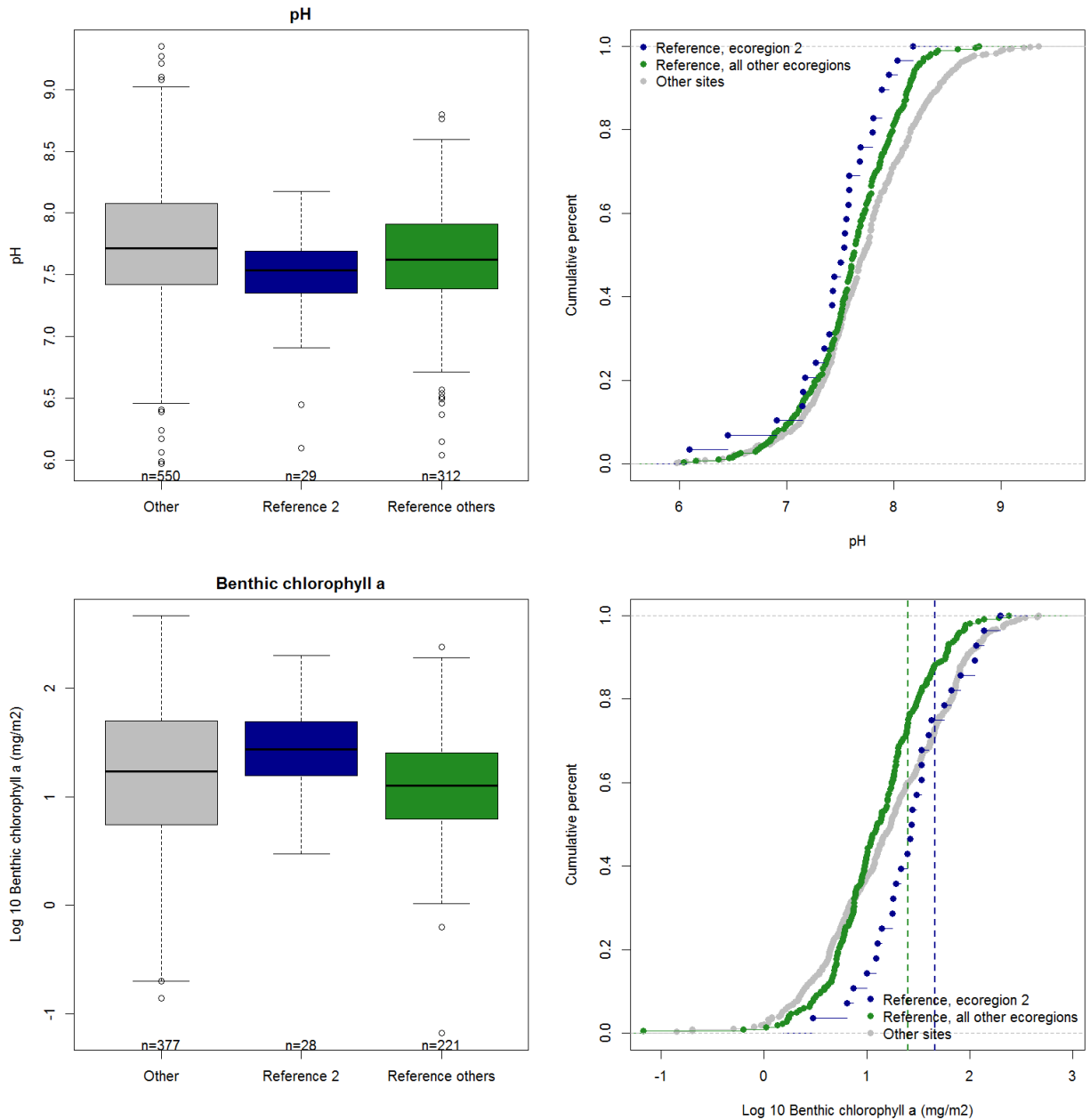


Figure 7 - Distributions of grab sample pH (top) and benthic chlorophyll a (bottom) in streams not specifically designated as reference (“Other”, grey symbols) and in reference streams, presented separately for Omernik Level III ecoregion 2 (blue symbols) and all other ecoregions (green symbols). The green dashed vertical line is the reference site 75th percentile for non-ecoregion 2 ecoregions. The dark blue dashed vertical lines is the 75th percentiles of ecoregion 2. Sample size for these figures are provided in the boxplots and also in Table 2.

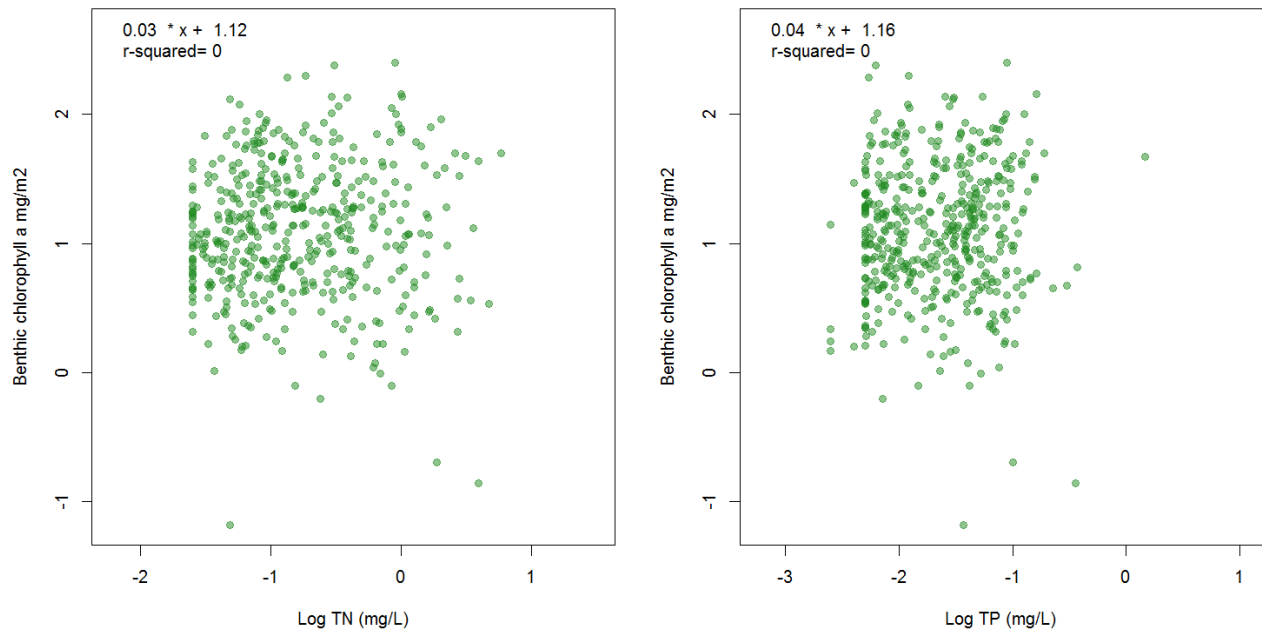


Figure 8 - Benthic chlorophyll *a* (mg/m²) in relation to grab sample nutrient concentration.

Table 3 - TN endpoints interpolated from periphyton metric regression models as responses to TN. All linear regressions presented were statistically significant ($p < 0.001$) in the proper response direction (ecologically sound). Rho = spearman's correlation coefficient, b = regression intercept, m = regression slope, r^2 = variance explained by regression, q = percentile of deciles (90th) and quartiles (75th), TN90 and TN75 are the interpolated log-base TN values first and back-transformed values second associated with the reference percentile of each metric. "wa" in front of a variable indicates the metric is based on a weighted-average tolerance value based approach. The metric types are explained in the Methods and Abbreviations of metrics and metric sources described in Appendix 2.

Metric name	rho	intercept	slope	r2	q90	TN90	q75	TN75		
wa_OptCat_LNtl	0.46	2.73	0.45	0.21	2.68	-0.11	0.78	2.28	-1.01	0.10
wa_OptCat_L1DisTot	0.45	2.90	0.45	0.19	2.81	-0.21	0.62	2.49	-0.91	0.12
wa_OptCat_DisTotMMI	0.38	2.52	0.48	0.14	2.57	0.10	1.26	2.15	-0.77	0.17
wa_OptCat_NutMMI	0.37	2.49	0.46	0.14	2.56	0.16	1.45	2.16	-0.71	0.19
wa_OptCat_XEMBED	0.36	2.24	0.42	0.15	2.15	-0.22	0.60	1.84	-0.94	0.11

Table 4 – TP endpoints interpolated from periphyton metric regression models as responses to TP. All other details as in Table 3.

Metric Name	rho	intercept	slope	r2	q90	TP90	q75	TP75		
wa_OptCat_L1PtI	0.67	3.31	0.72	0.41	2.52	-1.10	0.080	2.07	-1.72	0.019
wa_OptCat_NutMMI	0.65	3.36	0.71	0.38	2.57	-1.11	0.078	2.18	-1.66	0.022
wa_OptCat_DisTotMMI	0.63	3.34	0.69	0.34	2.61	-1.05	0.089	2.16	-1.71	0.020
wa_OptCat_XEMBED	0.58	2.86	0.56	0.30	2.15	-1.28	0.052	1.85	-1.81	0.015
wa_OptCat_LCond	0.57	3.18	0.61	0.27	2.58	-0.98	0.106	2.11	-1.74	0.018

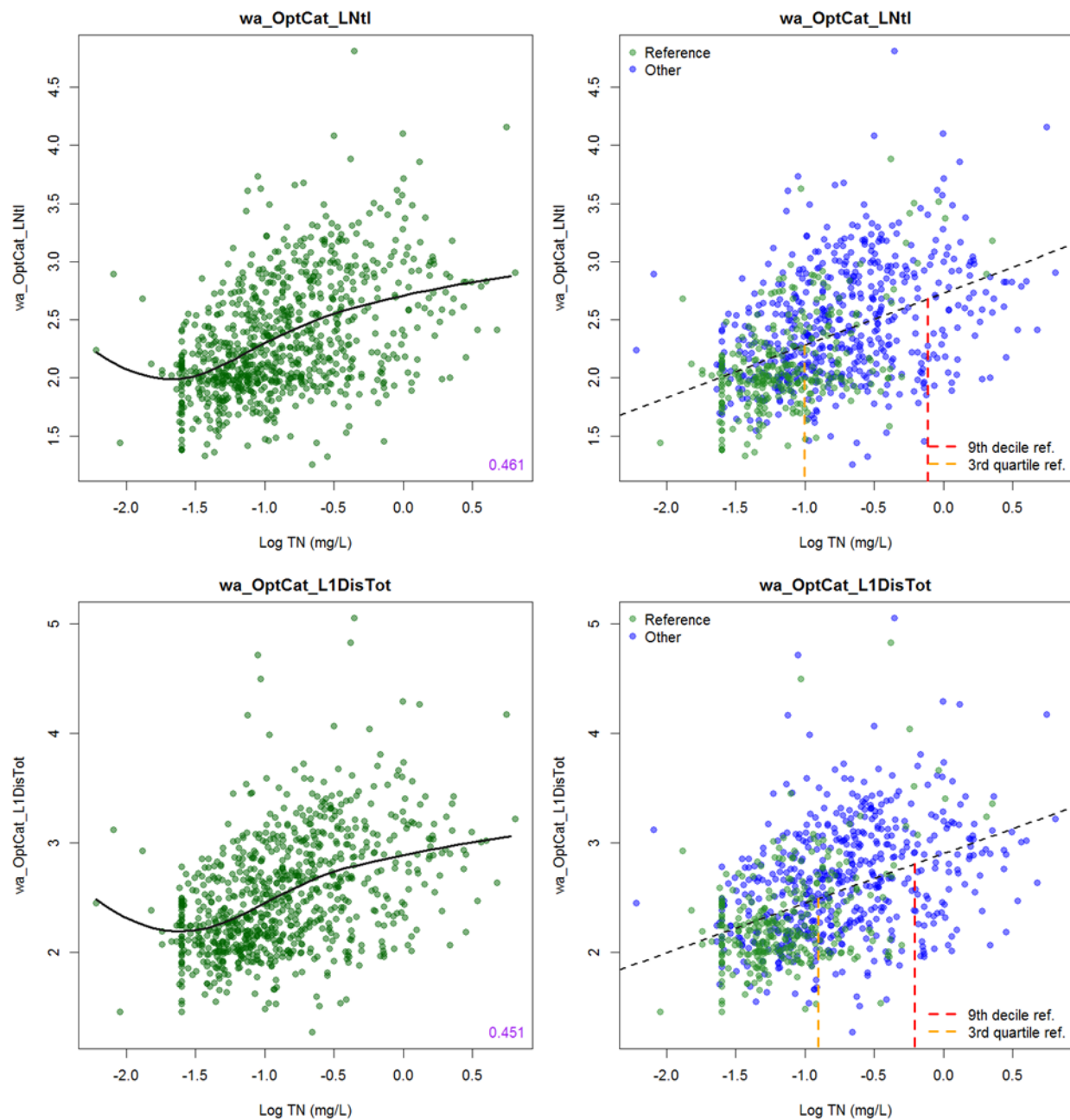


Figure 9 - Two selected diatom metrics shown in relation to TN. Loess lines plots (left) annotated with Spearman's rho (loess span=0.75). Simple linear regression model (right) with the 9th decile and 3rd quartile of the reference site metric distribution identified and the nutrient concentrations associated with those response targets represented as the dashed lines (right). For a description of the periphyton metrics, see Methods and Appendix 2.

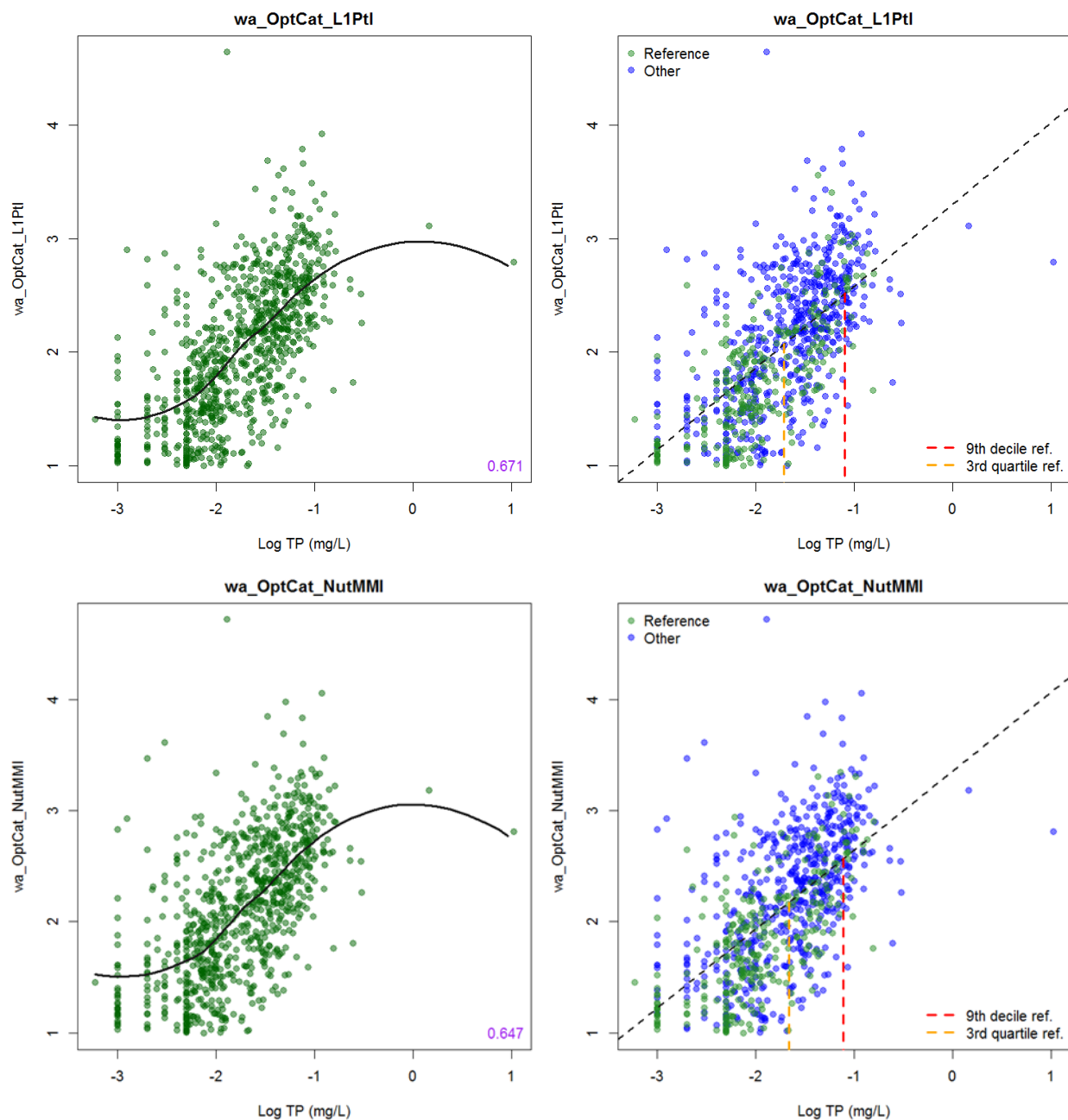


Figure 10 - Two selected diatom metrics shown in relation to TN. Loess lines plots (left) annotated with Spearman's rho (loess span=0.75). Simple linear regression model (right) with the 9th decile and 3rd quartile of the reference site metric distribution identified and the nutrient concentrations associated with those response targets represented as the dashed lines (right). For a description of the periphyton metrics, see Methods and Appendix 2.

Classification: *Residual analysis*

Residual analysis revealed that diatom metric - nutrient concentration relationships varied along gradients of latitude and precipitation (Figure 11 and 12; Figure 14 and 15). Variability was less along longitude and elevation gradients. Ecoregions 3, 10, and 11 tended to have higher residuals than other ecoregions for both nutrients, but particularly TN (Figure 13 and 16).

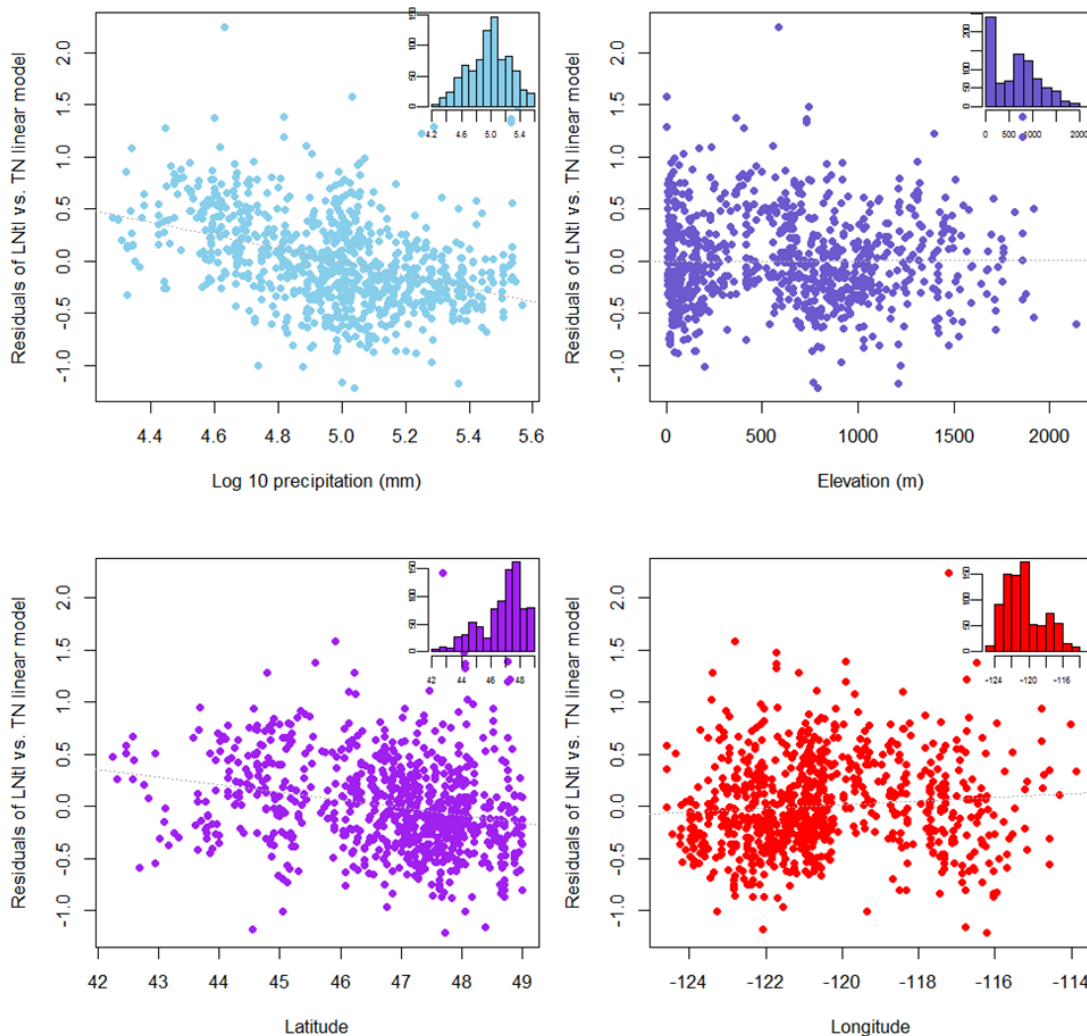


Figure 11 - Residuals of the weighted average LNTl diatom metric (wa_OptCat_LNTl) vs. log10 transformed TN linear model as a function of PRISM precipitation (top left), elevation (top right), latitude and longitude (bottom left and right, respectively).

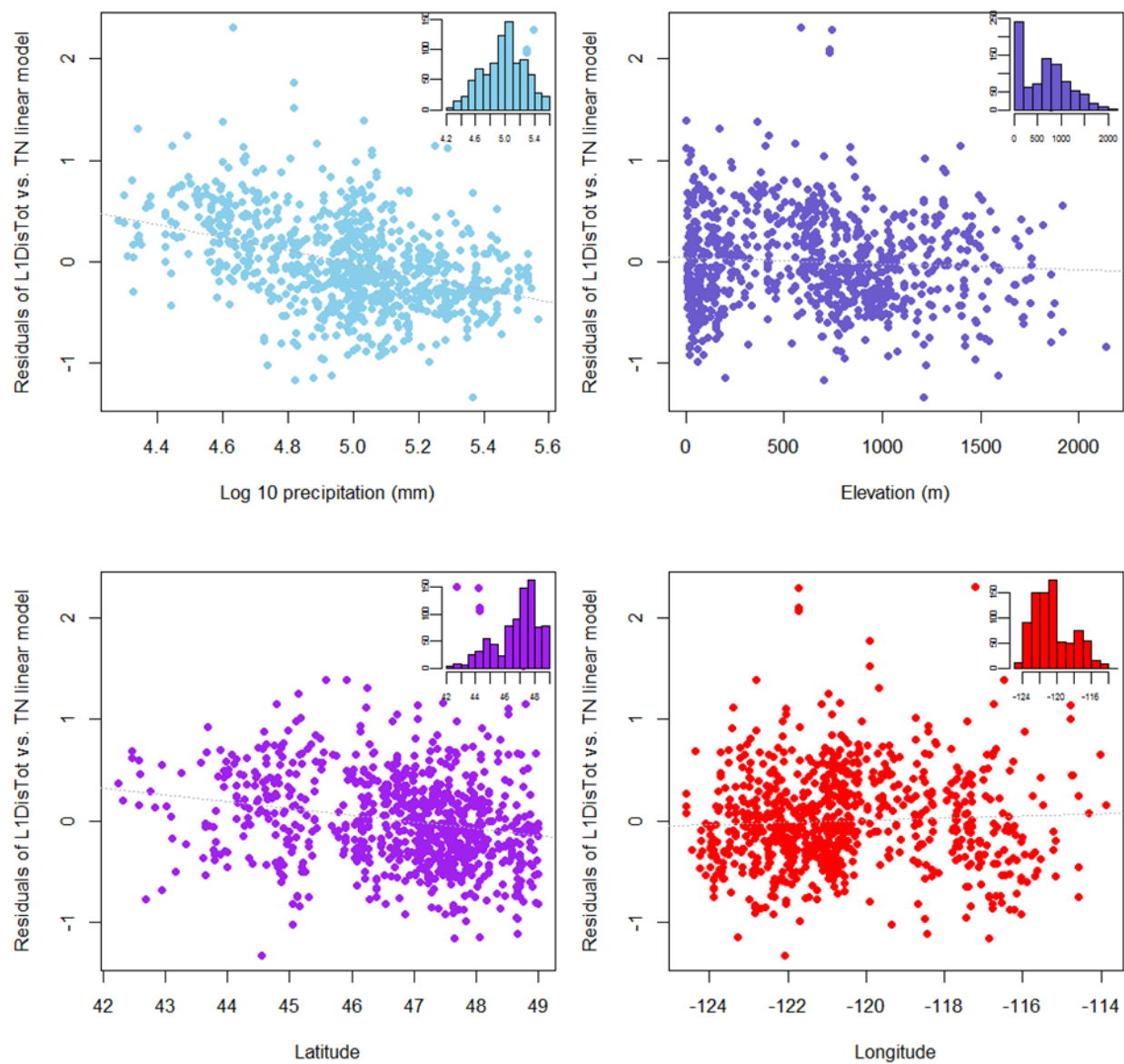


Figure 12 - Residuals of the weighted average L1DisTot diatom metric (wa_OptCat_L1DisTot) vs. \log_{10} transformed TN linear model as a function of PRISM precipitation (top left), elevation (top right), latitude and longitude (bottom left and right, respectively).

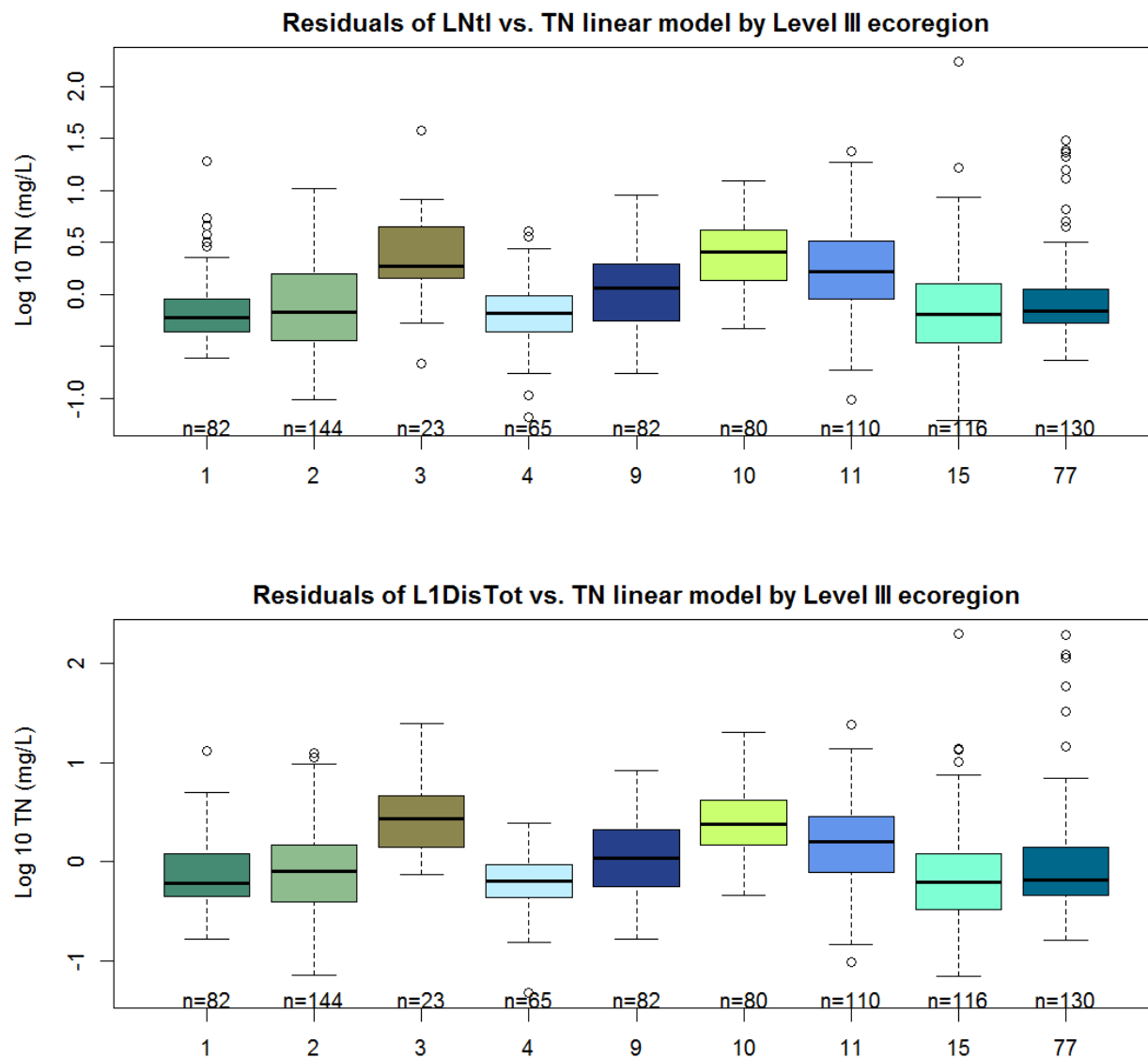


Figure 13 - Residuals of the weighted average LNTl diatom metric (wa_OptCat_LNTl) vs. \log_{10} transformed TN linear model (top) and weighted average L1DisTot diatom metric (wa_OptCat_L1DisTot) vs. \log_{10} transformed TN model (bottom) as a function of Omernik Level III ecoregion.

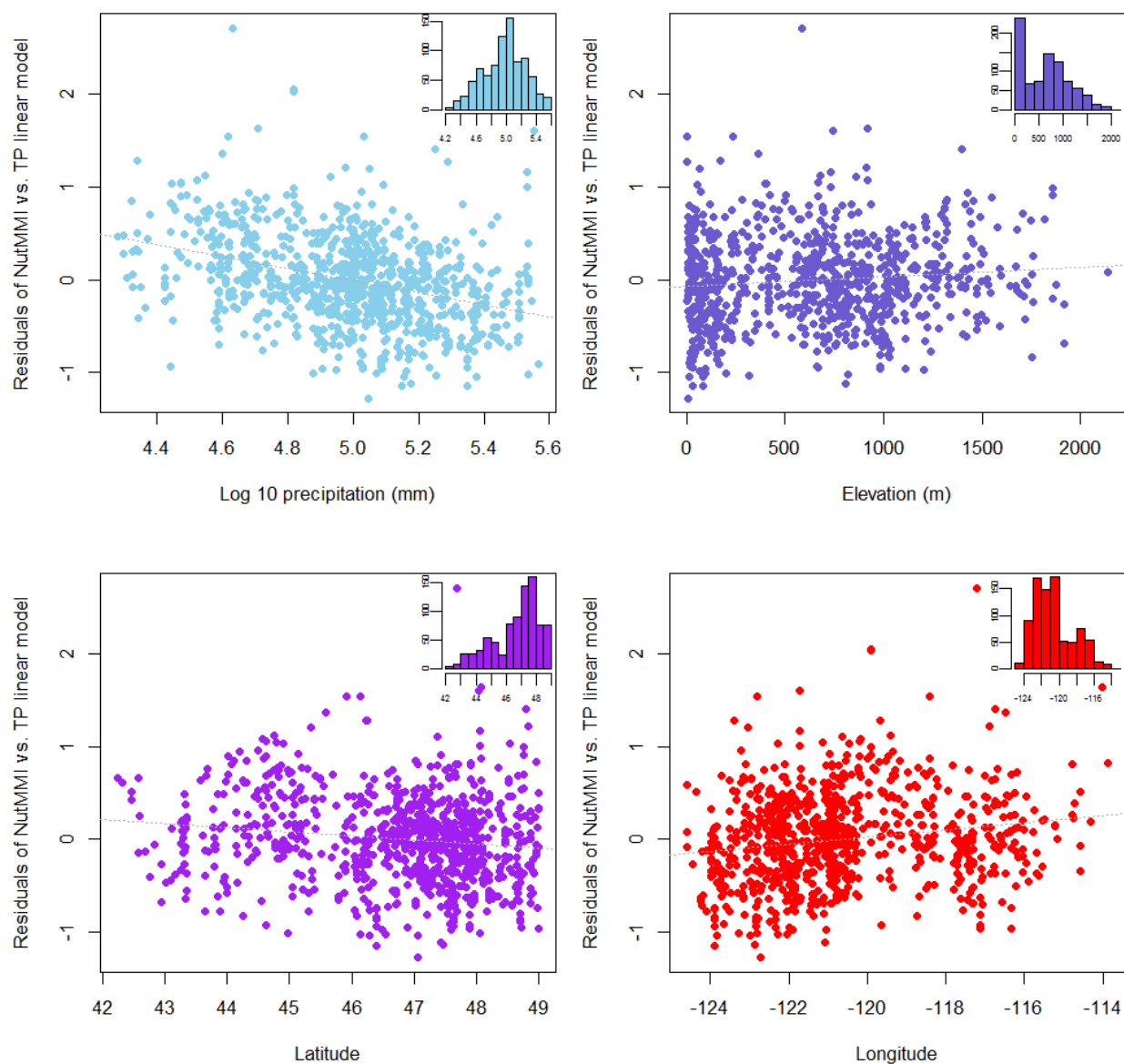


Figure 14 - Residuals of the weighted average NutMMI diatom metric (wa_OptCat_NutMMI) vs. \log_{10} transformed TN linear model as a function of PRISM precipitation (top left), elevation (top right), latitude and longitude (bottom left and right, respectively).

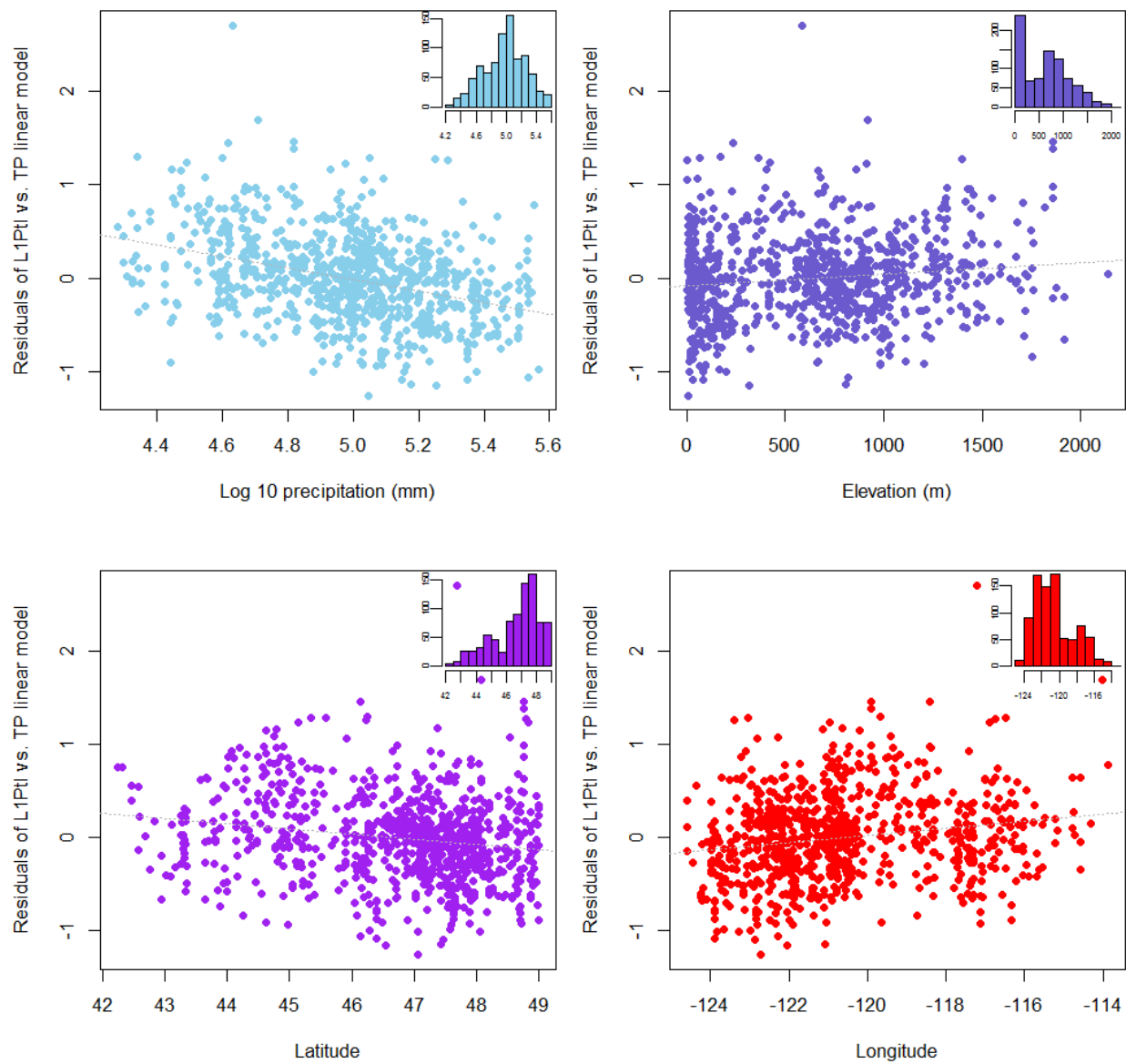


Figure 15 - Residuals of the weighted average L1Ptl diatom metric (wa_OptCat_L1Ptl) vs. log10 transformed TN linear model as a function of PRISM precipitation (top left), elevation (top right), latitude and longitude (bottom left and right, respectively).

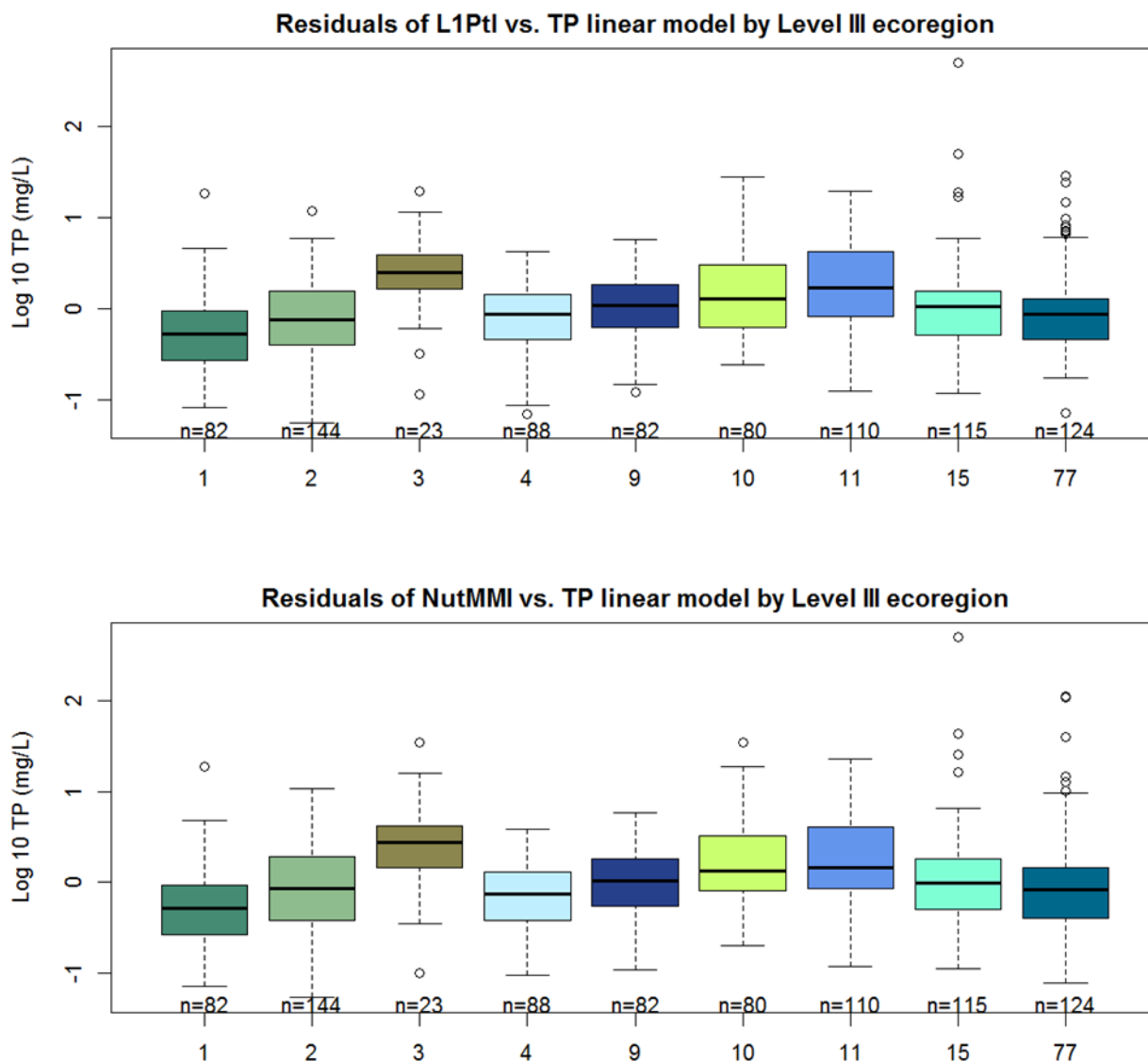


Figure 16 - Residuals of the weighted average L1Ptl diatom metric (wa_OptCat_L1Ptl) vs. log₁₀ transformed TP linear model (top) and weighted average NutMMI diatom metric (wa_OptCat_NutMMI) vs. log₁₀ transformed TP model (bottom) as a function of Omernik Level III ecoregion.

Model-based recursive partitioning

Model based-recursive partitioning for metrics responsive to TN indicated that latitude splits around 46.4° (South of Yakima) produced different models, with steeper slopes characteristic of lower latitudes (Figure 17). Model based-recursive partitioning for metrics responsive to TP produced latitude splits slightly higher, around 47.4° (between Tacoma and Seattle) (Figure 18). Splits in longitude for both nutrient models occurred between -121.2° to -121.8°. When ecoregion

was used as the splitting variable, models for both nutrients spilt with ecoregions 3, 10 and 11 apart from the rest (Figure 19 and 20). Additional model output shown in Appendix 4.

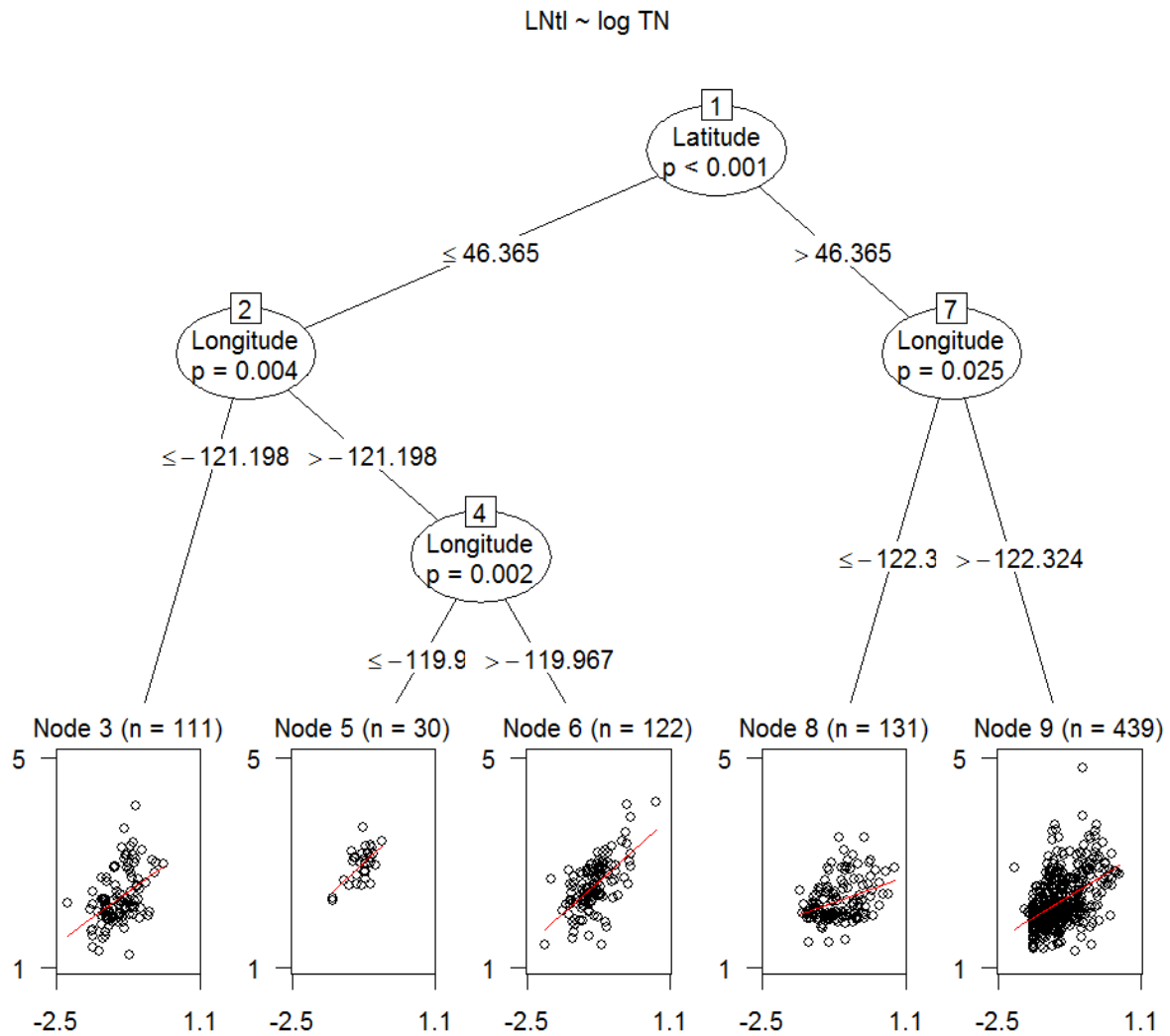


Figure 17 - Model-based recursive partitioning of the wa_OptCat_LNtl diatom optima metric as a response to TN concentration using longitude and latitude as potential splitting variables.

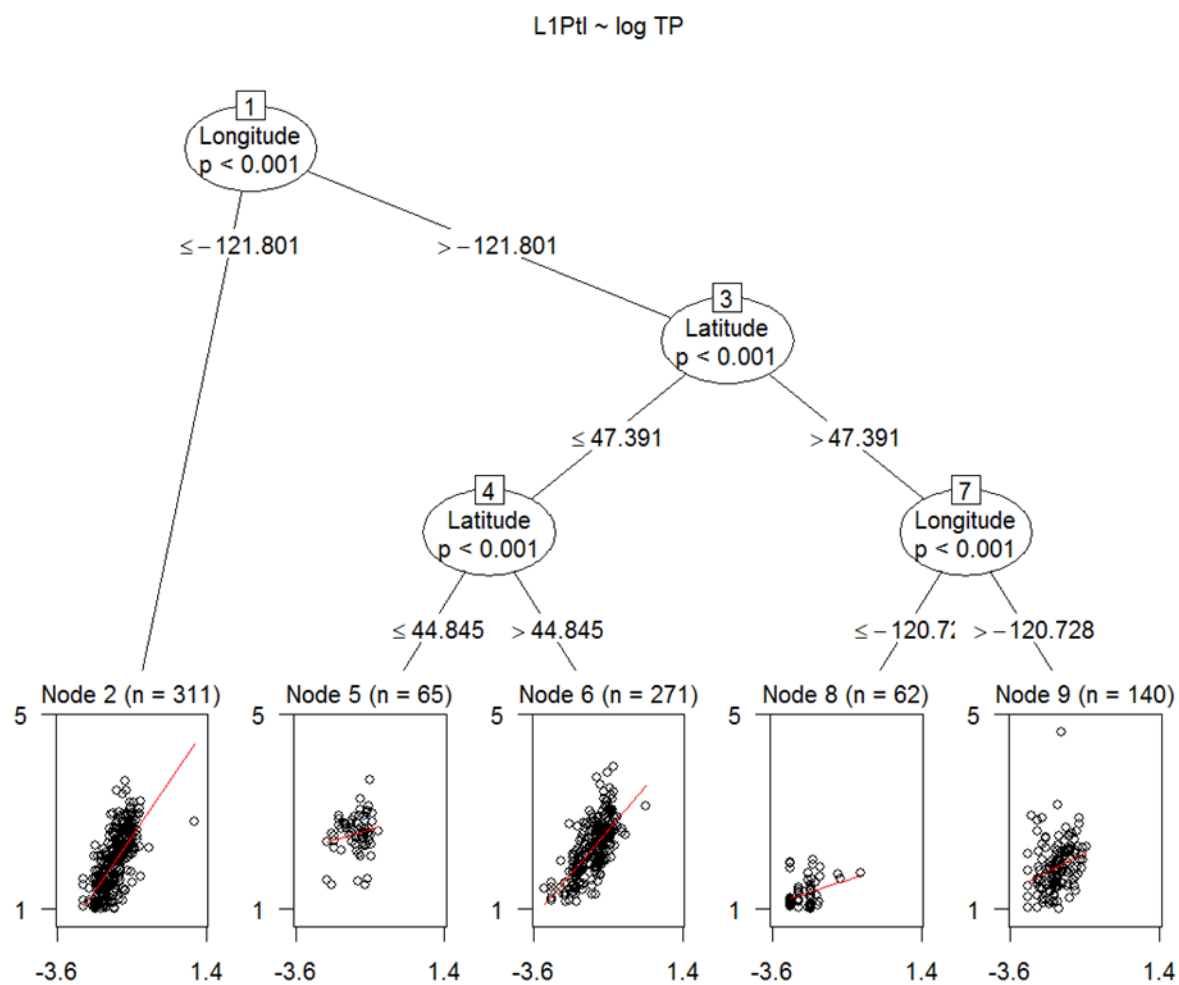


Figure 18 - Model-based recursive partitioning of the wa_OptCat_L1Ptl diatom optima metric as a response to TP concentration using longitude and latitude as potential splitting variables.

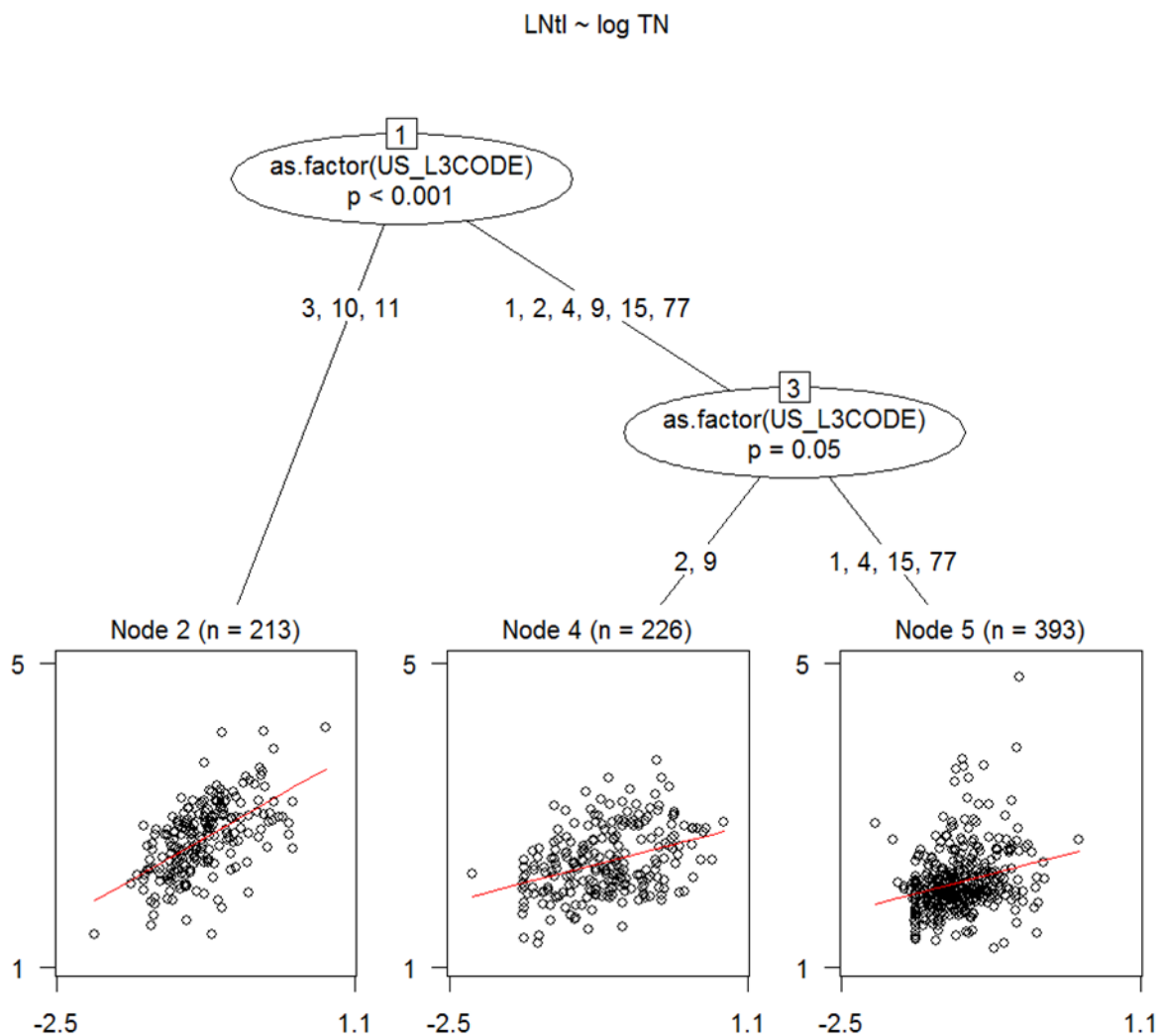


Figure 19 - Model-based recursive partitioning of the wa_OptCat_LNtl optima metric as a response to TN concentration using Omernik Level III ecoregion as the splitting variable.

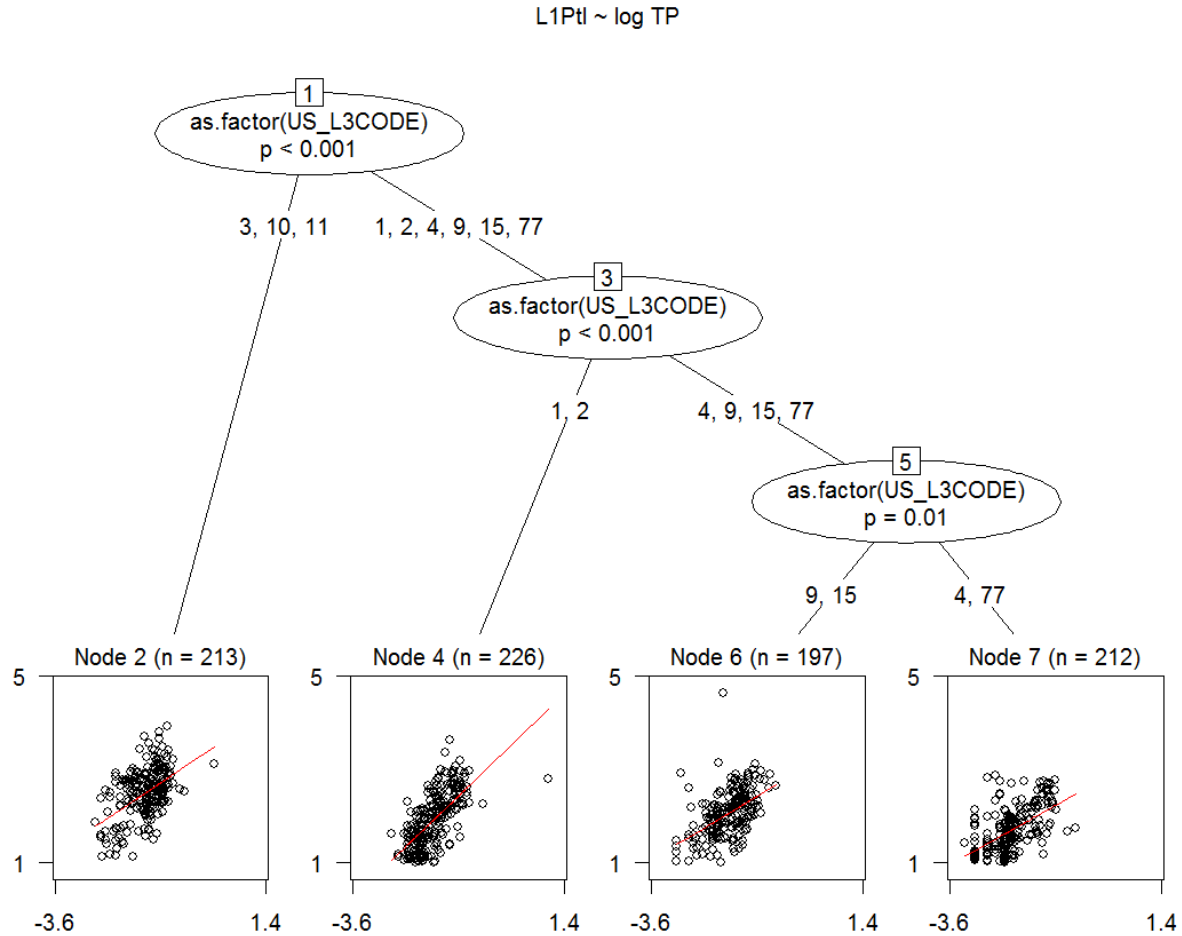


Figure 20 - Model-based recursive partitioning of the wa_OptCat_L1Ptl optima metric as a response to TP concentration using Omernik Level III ecoregion as the splitting variable.

Nash-Sutcliffe efficiency calculation

The NSE for models built from metrics sensitive to TN and log TN concentration using ecoregions 3, 10, and 11 had values around 0.34. The NSE values for the remaining data when compared to the ecoregion 3, 10 and 11 models were negative, indicating that the observed (test) data mean was a better estimate than the ecoregion 3, 10, and 11 model (Figure 21, top).

Models built from metrics sensitive to TP in response to log TP concentration showed a different pattern; the ecoregion 3, 10 and 11 models and lower NSE's (0.245 for OptCat_L1Ptl), and the remaining ecoregions had negative NSE's however, based on linear model fits, the TP models appeared to differ in intercept rather than slope (Figure 21, bottom).

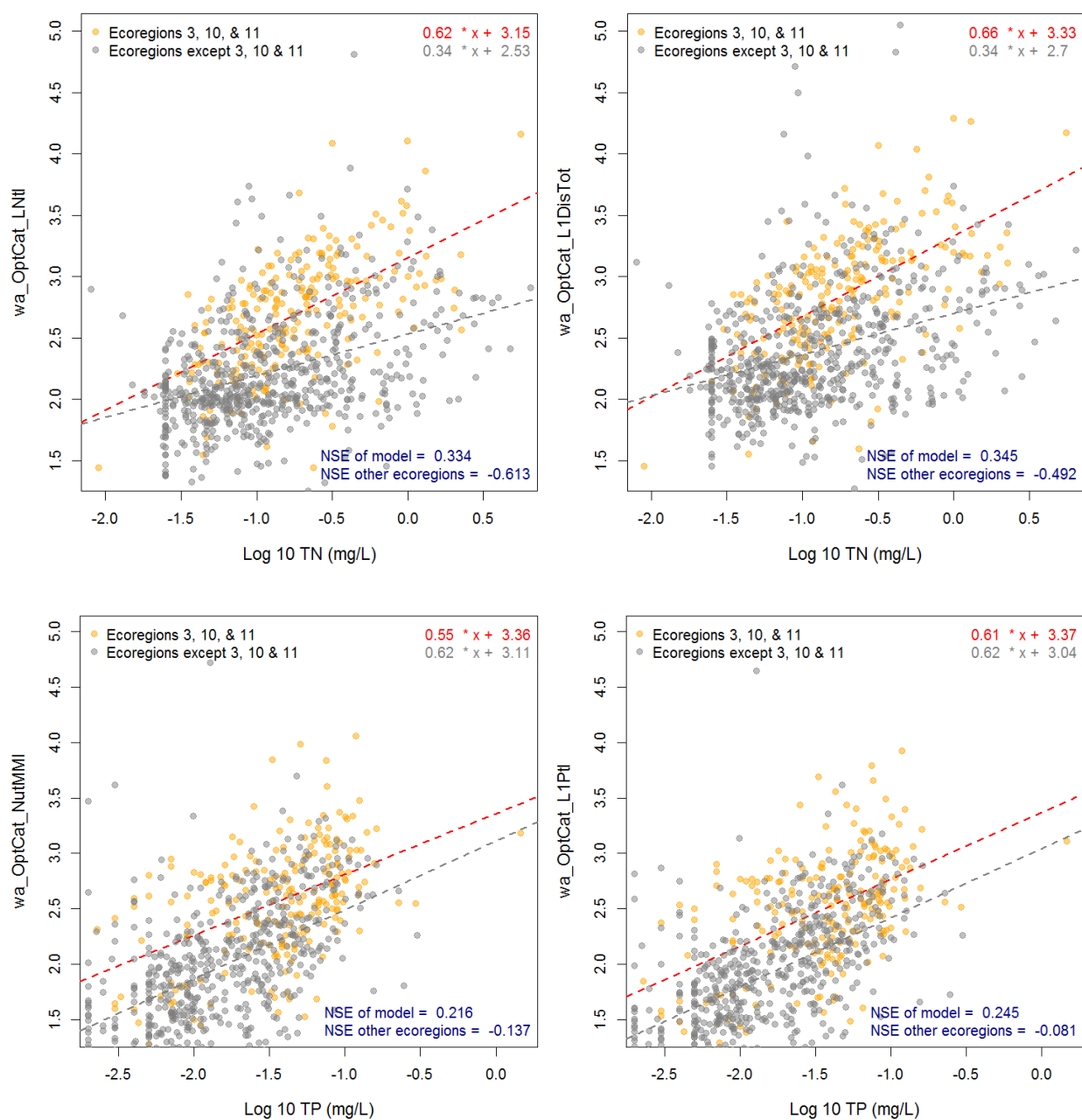


Figure 21 - Diatom metrics from sampling stations from ecoregions 3, 10, and 11 were used to build models against which models built from samples in all remaining ecoregions were tested. Top performing TN metrics (top) and TP metrics (bottom).

Stressor- response revisited – Ecoregional models

Based on the above classificatory lines of evidence, it was decided to separate ecoregions 3, 10 and 11 from the remaining Level III ecoregions in Washington State as two separate classes, and stressor-response models were rebuilt for these distinct groups. Although latitude also appeared to affect residuals, it did not have a uniform split point in the model-based recursive partitioning.

For TN, linear regression models for ecoregions 3, 10 and 11 have steeper slopes and higher intercepts than the remaining ecoregions (Figure 22 and 23, Table 5 and 6). For TP, the metrics in ecoregions 3, 10, and 11 has similar slopes but higher intercepts than the other ecoregions (Figure 24 and 25, Table 7 and 8).

Table 5 - TN endpoints interpolated from periphyton metric regression models as responses to TN in ecoregions 3, 10, and 11. All other details as in Table 3.

Metric name	rho	intercept	slope	r2	q90	TN90	q75	TN75		
wa_OptCat_L1DisTot	0.60	3.33	0.66	0.35	3.24	-0.13	0.74	2.95	-0.59	0.26
wa_OptCat_LNtl	0.58	3.15	0.62	0.33	2.99	-0.26	0.55	2.79	-0.58	0.27
wa_OptCat_DisTotMMI	0.53	3.04	0.62	0.27	2.97	-0.13	0.75	2.76	-0.45	0.35
wa_OptCat_NutMMI	0.51	2.98	0.56	0.23	2.95	-0.05	0.90	2.74	-0.41	0.39
wa_OptCat_XEMBED	0.39	2.54	0.48	0.18	2.58	0.08	1.19	2.36	-0.39	0.41

Table 6 - TN endpoints interpolated from periphyton metric regression models as responses to TN in ecoregions 1, 2, 4, 9, 15, and 77. All other details as in Table 3.

Metric name	rho	intercept	slope	r2	q90	TN90	q75	TN75		
wa_OptCat_LNtl	0.40	2.53	0.34	0.15	2.34	-0.57	0.27	2.13	-1.19	0.06
wa_OptCat_L1DisTot	0.38	2.70	0.34	0.13	2.56	-0.41	0.39	2.32	-1.13	0.07
wa_OptCat_DisTotMMI	0.29	2.26	0.35	0.09	2.20	-0.18	0.66	1.96	-0.86	0.14
wa_OptCat_NutMMI	0.28	2.24	0.34	0.09	2.22	-0.08	0.83	1.98	-0.78	0.17
wa_OptCat_XEMBED	0.28	2.08	0.35	0.11	1.97	-0.31	0.49	1.71	-1.06	0.09

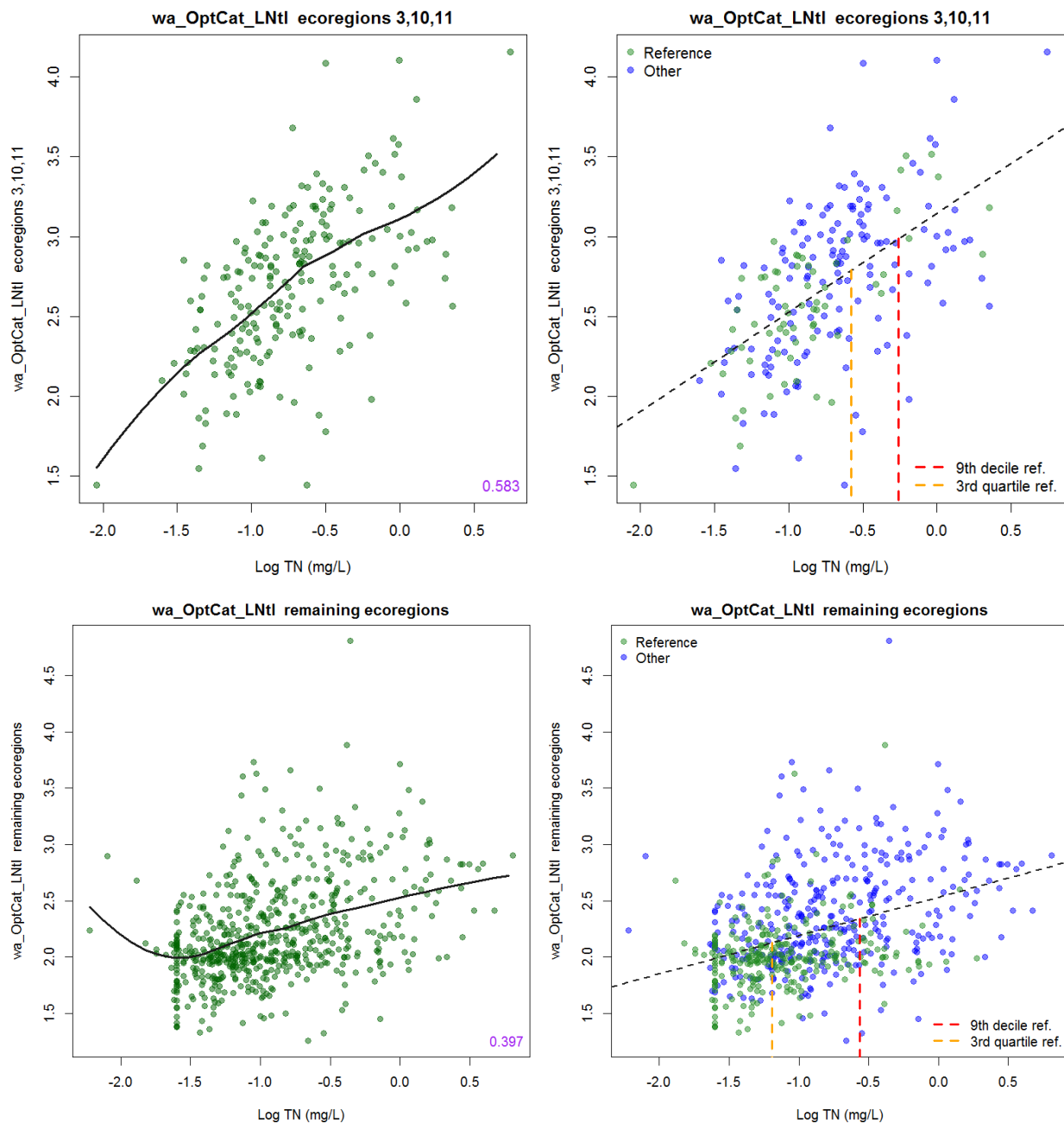


Figure 22 - Diatom metric wa_OptCat_L1Ntl in relation to TN in samples from ecoregions 3, 10 and 11 (top) and the remaining ecoregions (bottom). Loess lines plots (left) annotated with Spearman's rho (loess span=0.75). Simple linear regression model (right) with the 9th decile and 3rd quartile of the reference site metric distribution identified and the nutrient concentrations associated with those response targets represented as dashed lines (right). For a description of the periphyton metrics, see Methods and Appendix 2.

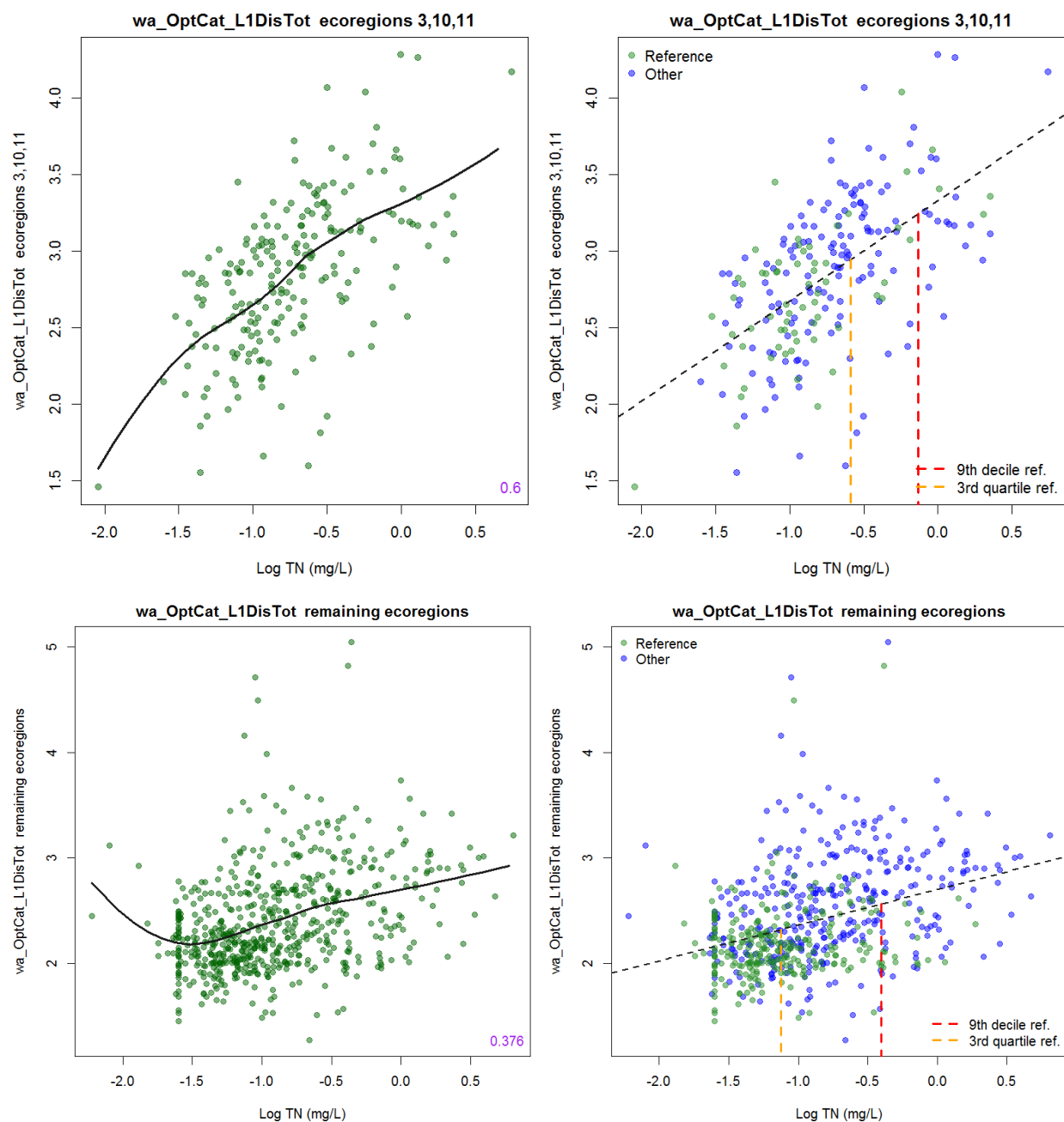


Figure 23 - Diatom metric wa_OptCat_DisTot in relation to TN in samples from ecoregions 3, 10 and 11 (top) and the remaining ecoregions (bottom). Loess lines plots (left) annotated with Spearman's rho (loess span=0.75). Simple linear regression model (right) with the 9th decile and 3rd quartile of the reference site metric distribution identified and the nutrient concentrations associated with those response targets represented as dashed lines (right). For a description of the periphyton metrics, see Methods and Appendix 2.

Table 7 - TP endpoints interpolated from periphyton metric regression models as responses to TP in ecoregions 3, 10, and 11. All other details as in Table 3.

Metric name	rho	intercept	slope	r2	q90	TP90	q75	TP75		
wa_OptCat_NutMMI	0.45	3.36	0.55	0.22	2.95	-0.75	0.18	2.74	-1.12	0.075
wa_OptCat_L1PtI	0.44	3.37	0.61	0.25	2.95	-0.69	0.20	2.68	-1.14	0.072
wa_OptCat_DisTotMMI	0.43	3.36	0.53	0.19	2.97	-0.74	0.18	2.76	-1.12	0.076
wa_OptCat_XEMBED	0.31	2.68	0.34	0.09	2.58	-0.31	0.49	2.36	-0.97	0.110
wa_OptCat_LCond	0.29	3.10	0.36	0.08	3.03	-0.19	0.65	2.72	-1.07	0.085

Table 8 - TP endpoints interpolated from periphyton metric regression models as responses to TP in ecoregions 1, 2, 4, 9, 15, and 77. All other details as in Table 3.

Metric name	rho	intercept	slope	r2	q90	TP90	q75	TP75		
wa_OptCat_L1PtI	0.65	3.07	0.63	0.37	2.18	-1.40	0.040	1.92	-1.81	0.016
wa_OptCat_NutMMI	0.62	3.13	0.63	0.33	2.22	-1.44	0.036	1.98	-1.82	0.015
wa_OptCat_DisTotMMI	0.60	3.09	0.61	0.30	2.21	-1.47	0.034	1.98	-1.85	0.014
wa_OptCat_XEMBED	0.57	2.78	0.54	0.29	1.97	-1.50	0.032	1.71	-1.97	0.011
wa_OptCat_LCond	0.55	2.92	0.53	0.24	2.20	-1.35	0.044	1.96	-1.82	0.015

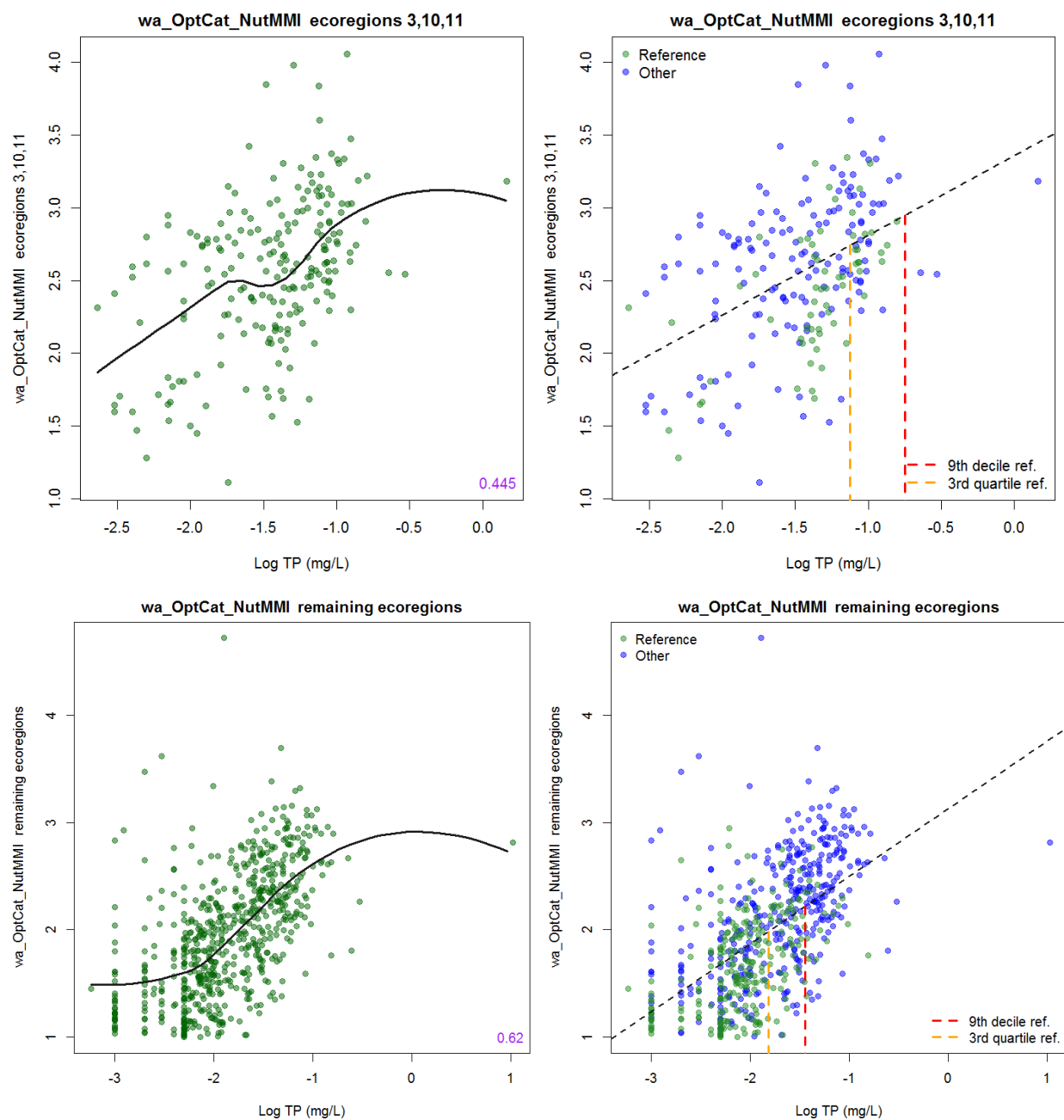


Figure 24 - Diatom metric *wa_OptCat_NutMMI* in relation to TP in samples from ecoregions 3, 10 and 11 (top) and the remaining ecoregions (bottom). Loess lines plots (left) annotated with Spearman's rho (loess span=0.75). Simple linear regression model (right) with the 9th decile and 3rd quartile of the reference site metric distribution identified and the nutrient concentrations associated with those response targets represented as dashed lines (right). For a description of the periphyton metrics, see Methods and Appendix 2.

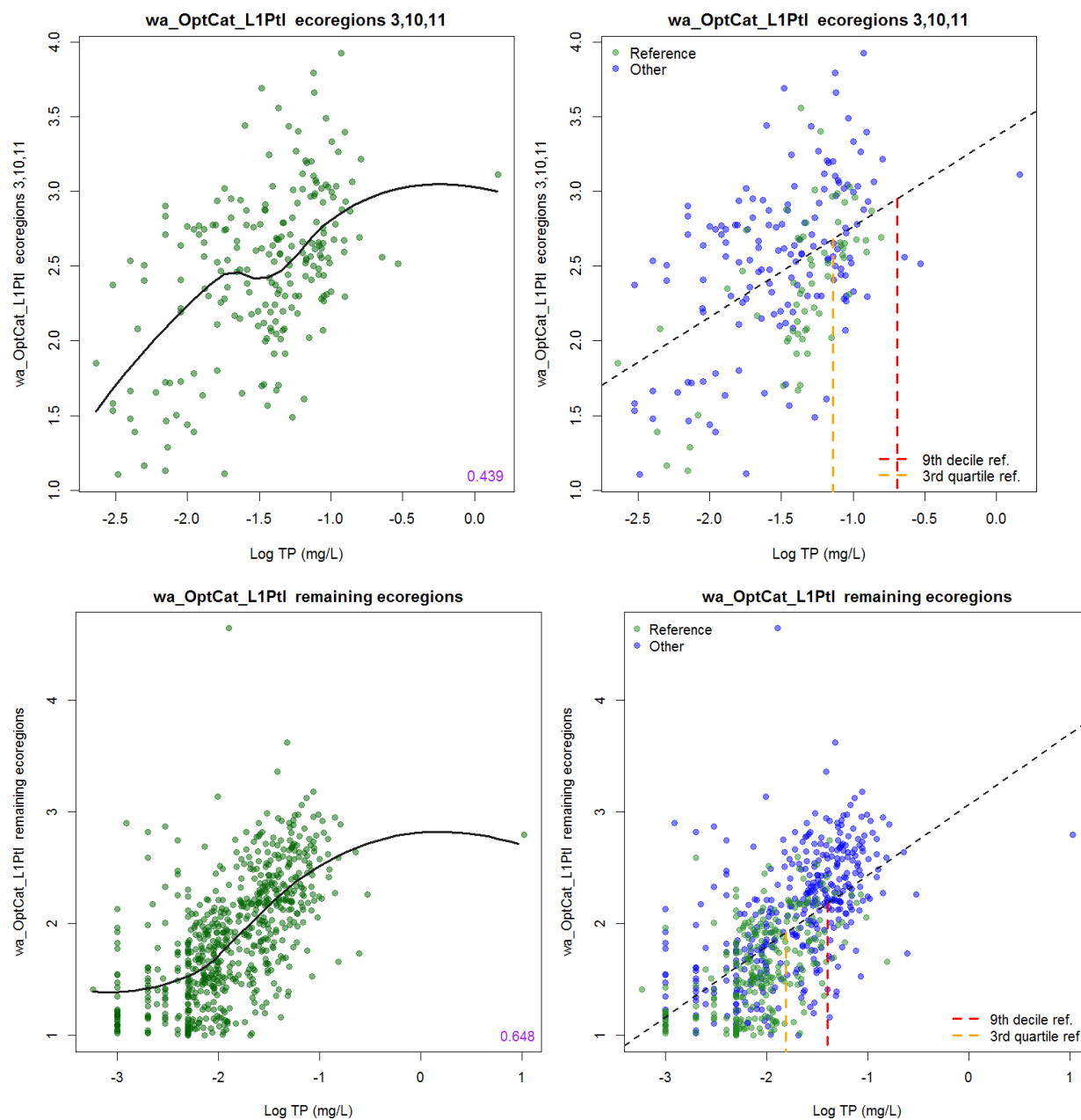


Figure 25 - Diatom metric wa_OptCat_L1Ptl in relation to TP in samples from ecoregions 3, 10 and 11 (top) and the remaining ecoregions (bottom). Loess lines plots (left) annotated with Spearman's rho (loess span=0.75). Simple linear regression model (right) with the 9th decile and 3rd quartile of the reference site metric distribution identified and the nutrient concentrations associated with those response targets represented as dashed lines (right). For a description of the periphyton metrics, see Methods and Appendix 2.

4. DISCUSSION

General Observations

Periphyton metrics responded to nutrient concentration gradients in Washington. As suggested before in many contexts, diatom assemblage structure is a sensitive indicator of nutrient gradients. This is likely not surprising but it is important to demonstrate given the sometimes confusing anecdotes that one hears on algae not being sensitive to nutrients in streams. Algal composition is especially sensitive to nutrients in Washington streams. Algal biomass is hard to accurately characterize at the reach scale and often shows highly variable if not insignificant responses to nutrients. Indeed, in Washington, chlorophyll *a* showed little directional response to nutrient gradients. There are many reasons for this, methodological and ecological. However, the algal assemblage composition data exhibited a consistent and significant response to both total N and P.

The significance of this observation is that algal assemblage tools have now demonstrated their potential for a number of applications in Washington. The responsiveness to nutrients, especially above reference conditions, means they can be used, at a minimum, as screening tools or confirmatory tools of nutrient pollution and potentially other related water quality impairments (e.g., dissolved oxygen, pH, etc.). Either the values of nutrients derived from the stressor-response analyses above, or the reference based periphyton metric values could both be considered in this capacity.

This analysis also supports the application of algal assemblage structure as a potential additional biological indicator for aquatic life use. Algae are, of course, aquatic life, and EPA recommends their use in biological assessment, along with macroinvertebrates and fish, to provide a complete assessment of biological condition (USEPA 2011). A primary requirement of any biological indicator is that it be responsive to disturbance. This study provides evidence that algae respond to nutrient disturbance; and there are values across the range of reference and non-reference conditions, indicating utility for resolving a wide range of nutrient pollution conditions.

Another noteworthy observation was the quality of the stressor-response relationships. Most all of the periphyton metric – nutrient response relationships were significant. Also, for many stressor-response relationship models, nutrient concentrations explained more than a third of the variability in metric scores, which is relatively high in comparison to the chlorophyll *a* – nutrient relationships or AFDM – nutrient relationships, and similar to that observed in other studies of periphyton metrics (e.g., Stevenson et al. 2008).

Application

A number of potential applications are recommended by this analysis. First, the cumulative distribution of endpoint values derived from the stressor-response relationship models provide reasonable TN and TP screening values consistent with the protection of reference quality periphyton conditions in streams in Washington, as do the periphyton metric reference percentiles themselves. These nutrient distributions, though, indicate a wide range of nutrient endpoints associated with different responses from very sensitive periphyton indicators, with alterations outside of the reference condition occurring as low as 0.010 mg/L TP and 0.100 mg/L TN, to less sensitive indicators with much higher nutrient values. WDOE will need to decide where along these distributions a screening value should be located. The eye is naturally drawn to a central tendency, but this should be weighed against whether an impact on 50% of periphyton measures is sufficiently protective. Indeed, much of the distribution tails may be due to variability; but where does sufficient protection occur? In either case, some exploration of the individual metrics themselves is warranted and a combination of

statistical and ecological reasoning should be applied (e.g., are there specific algal attributes of greater concern).

These periphyton metrics are likely not ready to be used as standalone biological indicators, even though many come from peer-reviewed products and many are well established (i.e., akin to the EPT richness score for macroinvertebrates). There would still need to be several steps taken to assure their thoroughness and defensibility for such an application. However, they show great promise in this regard for Washington. Washington ECY adding another assemblage (algae) for routine monitoring to characterize biological integrity adds greater depth and breadth to assessment of stressors, and other entities throughout the state are encouraged to sample additional assemblages (USEPA 2011). Outside of assessing aquatic life use, however, the metrics may be sufficiently defensible for evaluating reasonable potential, TMDL monitoring and modeling, etc.

Recommended Additional Analyses

There are several next steps WDOE could take and the following are just a few suggestions. First, revisiting the classification is recommended. Algal assemblages could be structured by some of the common gradients influential in constraining species distributions – like temperature, pH, flow, etc. We just observed little structure with regards to these drivers, but it is recommended that this be pursued somewhat further possibly using reduced species lists (i.e., removing rare and common taxa).

Testing the relationships observed here with independent data, either collected specifically to test the gradients or by accessing additional periphyton data, is also recommended; however, this analysis represents the third application of this approach regionally, with comparable results. While this analysis encompassed several different programs, a broad nutrient gradient, and is likely robust, testing the responses in subwatersheds, local regions, or along known nutrient gradients within a small region may help strengthen confidence in the applicability.

Another additional effort that could be considered is an attempt to tie periphyton metric values to other valued ecological endpoints, such as nuisance algal levels (e.g., as defined with user perception studies), dissolved oxygen conditions, or invertebrate conditions. Certainly, one would expect periphyton to influence other assemblages, but linking different levels of periphyton metric values to levels of invertebrate responses would potentially help strengthen linkages to higher trophic levels. Similarly, dissolved oxygen concentrations and diel swings are pathways through which excess algal biomass is known to affect aquatic ecosystems. The degree to which linkages between DO and algal metrics could be made would also help strengthen this linkage and help screen sites for potential impairments and verify TMDL modeling of pollutant sources. Some metrics are designed for DO sensitivity, so there is a presumed linkage, but this could be tested in Washington as well.

A last step to be considered is inference modeling. Many diatoms have had nutrient optima calculated for them and/or that effort could be done for taxa in Washington. Once identified, transfer coefficients can be used to infer the true average nutrient concentrations present at a particular site. Indeed, this has been the approach New Jersey has used for their diatom index (Ponader et al. 2008). Quite often, grab samples or even averages from baseflow grab samples, are poor representations of average nutrient conditions. Diatoms may provide a more accurate picture. This may be especially useful in developing or monitoring targets.

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APPENDIX 1. Water quality variables used.

Table 9 -Variable names combined and method codes excluded. Abbreviations as in Table 2.

Variable name	Original variable name(s)	Method codes	Description
AFDM	AFDM, AFDM_mgm2, Biomass, periphyton, ash free dry mass, grams per square meter	All	Ash-free dry mass
Chla	Chlorophyll a, periphyton, chromatographic-fluorom; Chla; Chla_mgm2	Excluded 89	Chlorophyll a of periphyton in mg/m2
Cond	Cond_uScm, CondCal_uScm, CondHO_uScm		Specific conductivity in μ S/cm
DO_mgL	DO_mgL		Dissolved oxygen, mg/L
NH4_f		CL035, 37, 39, SHC02 (excluded 101)	Ammonium, filtered
NH4_uf	NH4_N_mgL	Excluded CL075	Ammonium, unfiltered, in mg/L as nitrogen
NO2_f	NO2_N_f_mgL	Excluded 49	Nitrite, filtered in mg/L as nitrogen
NO2_uf	NO2_N	-999, CL076, 77, 135	Nitrite, unfiltered in mg/L as nitrogen
NO3_f		All	Nitrate, filtered in mg/L as nitrogen
NO3_uf	NO3_N_mgL	-999 and ALGOR	Nitrate, unfiltered in mg/L as nitrogen
NOx_f		CLO45, 48, 50, 44, 47, 49, -999, CDR06, (excluded CL132, RED01,02)	Nitrate plus nitrite, filtered, in mg/L as nitrogen
NOx_uf	NOx_N	-999, 81, (excluded CL131)	Nitrate plus nitrite, unfiltered, in mg/L as nitrogen
OrgN_f	TKN (parameter codes 706 and 6230)	Excluded CL061	Organic nitrogen, filtered in mg/L
OrgN_uf	TKN_N (parameter codes 625 and 605)		Organic nitrogen, unfiltered in mg/L
OrthoP_uf	Orthophosphate, water, unfiltered, milligrams per liter		Orthophosphate, unfiltered in mg/L
SRP	SRP_mgL; Orthophosphate, water, filtered, mg per liter		Soluble reactive phosphorus in mg/L
TempC	Temp_degC		Temperature in $^{\circ}$ C
TN_f	TN_f_mgL (parameter codes 62854 and 6020)		Total nitrogen, filtered in mg/L
TN_uf	TN_N_mgL	600, 62855, 71887	Total nitrogen, unfiltered in mg/L
TP_uf	P_P_mgL, TP_P_mgL	Excluded CL84, 90, and 001	Total phosphorus, unfiltered in mg/L
TP_f	TDP_mgL	Excluded CL52 and 60	Total phosphorus, filtered in mg/L

APPENDIX 2. Diatom metrics used.

Table 10 - List of periphyton metrics used, type (continuous or categorical) and value range, description, expected response direction to nutrient enrichment, and sources.

Field	Type, values	Description	Response to nutrients	Source
wa_OptCat_DisTotMMI	Continuous, (-40 to 70)	Weighted average multivariate optima (raw score)	+	1
wa_OptCat_L1DisTot	Continuous, 0-4+	Diatom optima values to watershed disturbance, ranging from 0-4	+	1
wa_OptCat_L1Ptl	Continuous, 1-6+	Diatom optima values to log(TP), ranged from 0-6	+	1
wa_OptCat_LNtl	Continuous, 4-9	Diatom optima values to log(TN), ranged between 3-8	+	1
wa_OptCat_LCond	Continuous, 3-8	Diatom optima values to conductivity, ranged between 3-8	+	1
wa_OptCat_NutMMI	Continuous, (-40 to 70+)	Weighted average nutrient multivariate optima (raw score)	+	1
wa_OptCat_pH	Continuous, (7.1 to 8.6, pH)	Diatom optima values to pH, ranged between 7.1-8.6	+	1
wa_OptCat_XEMBED	Continuous, (0-100)	Diatom optima values to embeddedness ranged from 0-100	+	1
wa_OptCat_PctFN	Continuous, (0-5, actual pct 0-100)?	Diatom optima values to percent fine ranged from 0-100	+	1
NEWTSC	Continuous, 1-6	MAIA/MAHA assessment result, Trophic State Index from oligotrophic (1) to hypertrophic (6) state	+	2
MAIATSC	Continuous, 1 to 6.25	MAIA/MAHA assessment result, Trophic State Index from oligotrophic (1) to hypertrophic (6) state	+	2
Ptpv_TP_all_Ind	Categories, text	USGS NAWQA program TP indicator values for the entire dataset (1 or 2); 1 is high TP, 2 is low TP		4
Ptpv_TN_all_Ind	Categories, text	USGS NAWQA program TN indicator values for the entire dataset (1 or 2); 1 is high TN, 2 is low TN		4
Ptpv_TP_WM_Ind	Categories, text	USGS NAWQA program TP indicator values for Western Mountain (WM); 1 is high TP, 2 is low TP		4
Ptpv_TN_WM_Ind	Categories, text	USGS NAWQA program TN indicator values for Western Mountain (WM); 1 is high TN, 2 is low TN		4
Ptpv_TP_CWP_Ind	Categories, text	USGS NAWQA program TP indicator values for central, western plains (CWP); 1 is high TP, 2 is low TP		4
Ptpv_TN_CWP_Ind	Categories, text	USGS NAWQA program TN indicator values for central, western plains (CWP); 1 is high TN, 2 is low TN		4
POLL_CLASS	Categories, 1 to 4	From pollution tolerant taxa (1) to pollution sensitive taxa (4)	-	7
POLL_TOL	Categories, 1 to 5	From pollution tolerant taxa (1) to pollution sensitive taxa (5)	-	8
BEN_SES	Categories, 1 or 2	Sestonic (2) or benthic algae (1)	+	7
DIATAS_TP	Categories, 1 or 2	Diatom TP preference (1)		7
DIATAS_TN	Categories, 1 or 2	Diatom TN preference (1)		7

Sources:

1 = Stevenson, R. J., Y. Pan, K. M. Manoylov, C. A. Parker, D. P. Larsen, and A. T. Herlihy 2008. Development of diatom indicators of ecological conditions for streams of the western US. J. N. Am. Benthol. Soc., 2008,

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APPENDIX 3. Additional diatom metrics shown in relation to nutrient concentrations.

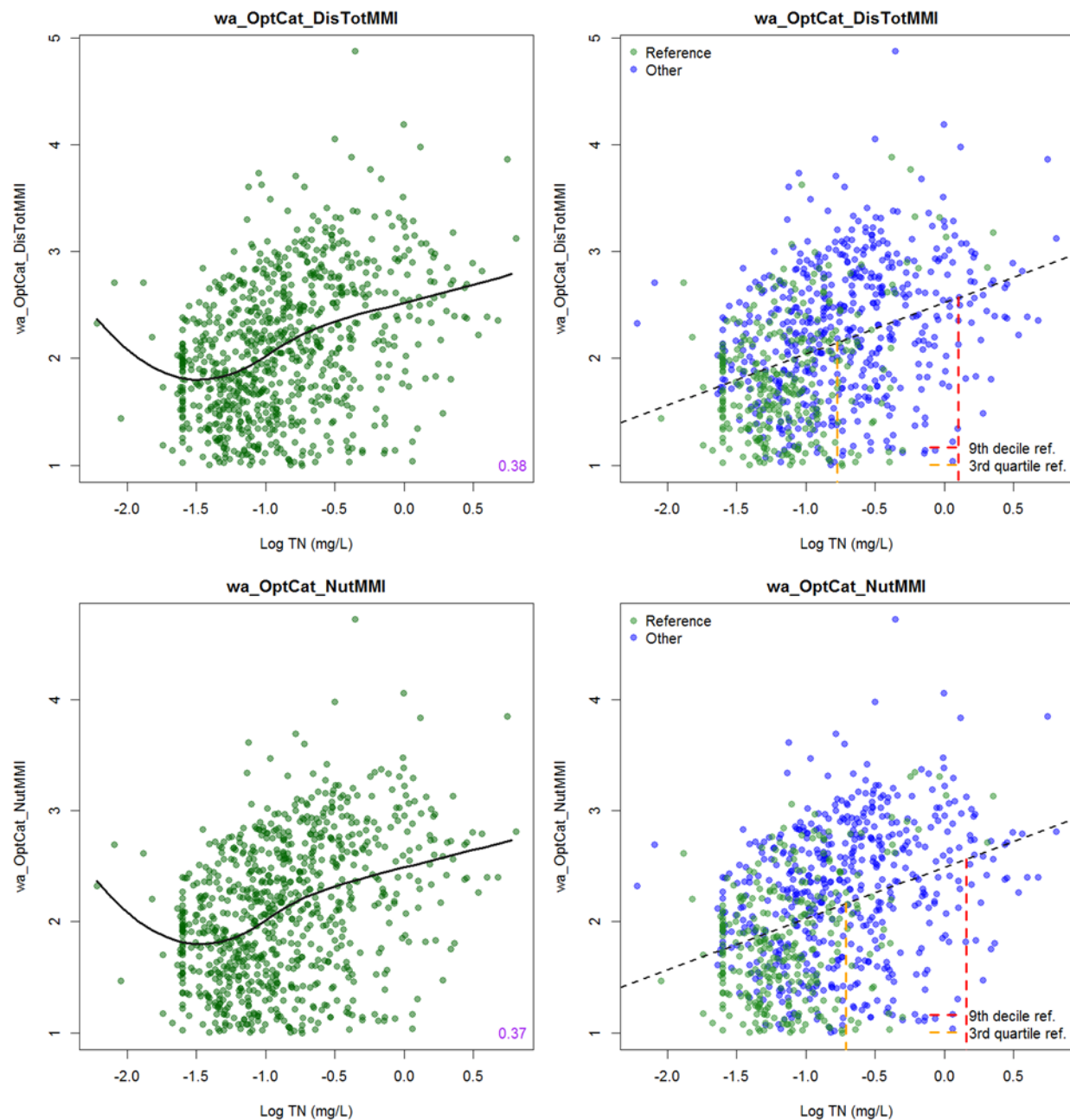


Figure 26 - Loess lines plots (left) annotated with Spearman's rho (loess span=0.75). Simple linear regression model (right) with the 9th decile and 3rd quartile of the reference site metric distribution identified and the nutrient concentrations associated with those response targets represented as the dashed lines (right). For a description of the periphyton metrics, see Methods and Appendix 2.

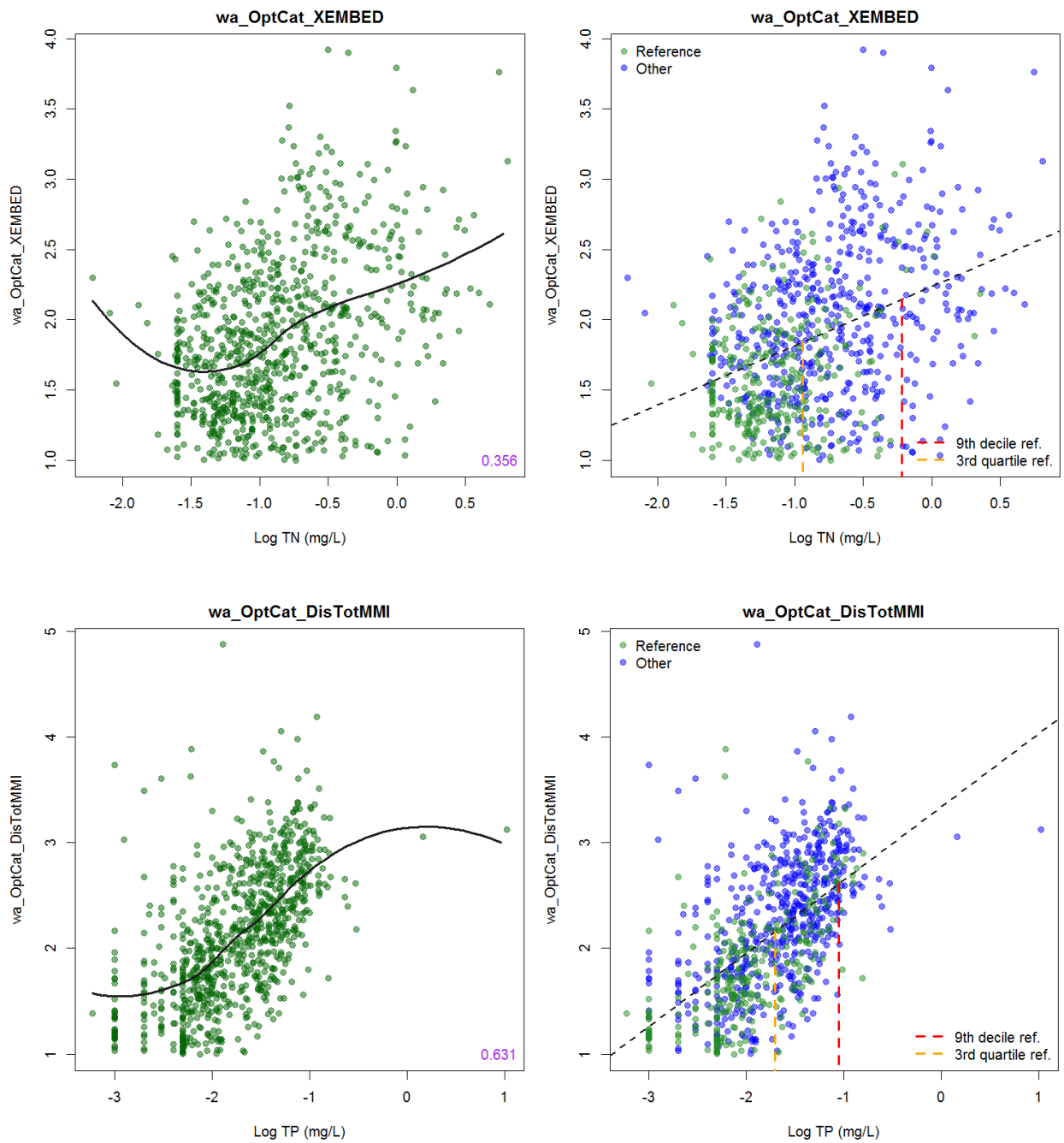


Figure 27 - Loess lines plots (left) annotated with Spearman's rho (loess span=0.75). Simple linear regression model (right) with the 9th decile and 3rd quartile of the reference site metric distribution identified and the nutrient concentrations associated with those response targets represented as the dashed lines (right). For a description of the periphyton metrics, see Methods and Appendix 2.

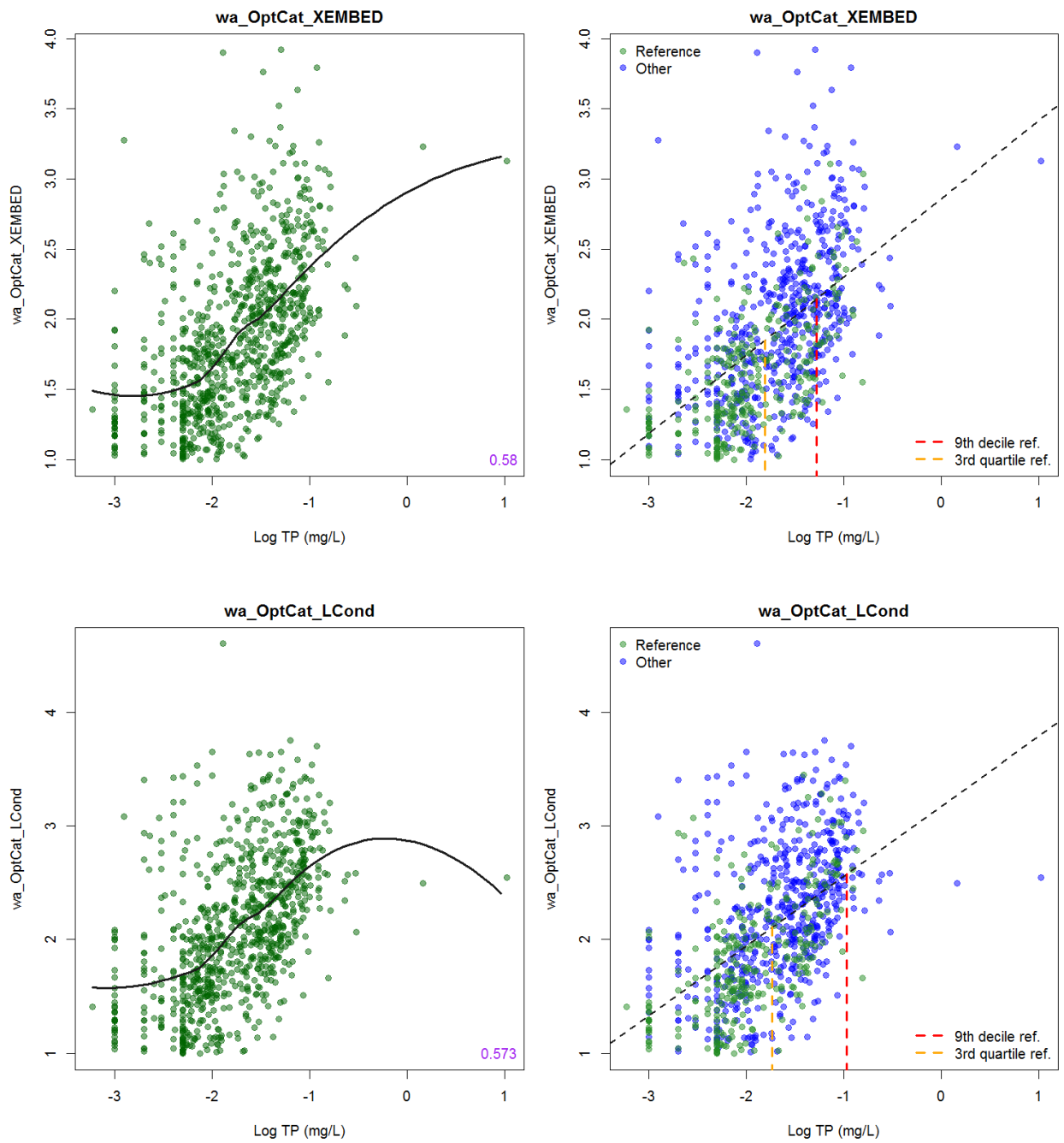


Figure 28 - Loess lines plots (left) annotated with Spearman's rho (loess span=0.75). Simple linear regression model (right) with the 9th decile and 3rd quartile of the reference site metric distribution identified and the nutrient concentrations associated with those response targets represented as the dashed lines (right). For a description of the periphyton metrics, see Methods and Appendix 2.

APPENDIX 4. Supplementary model-based recursive partitioning figures.

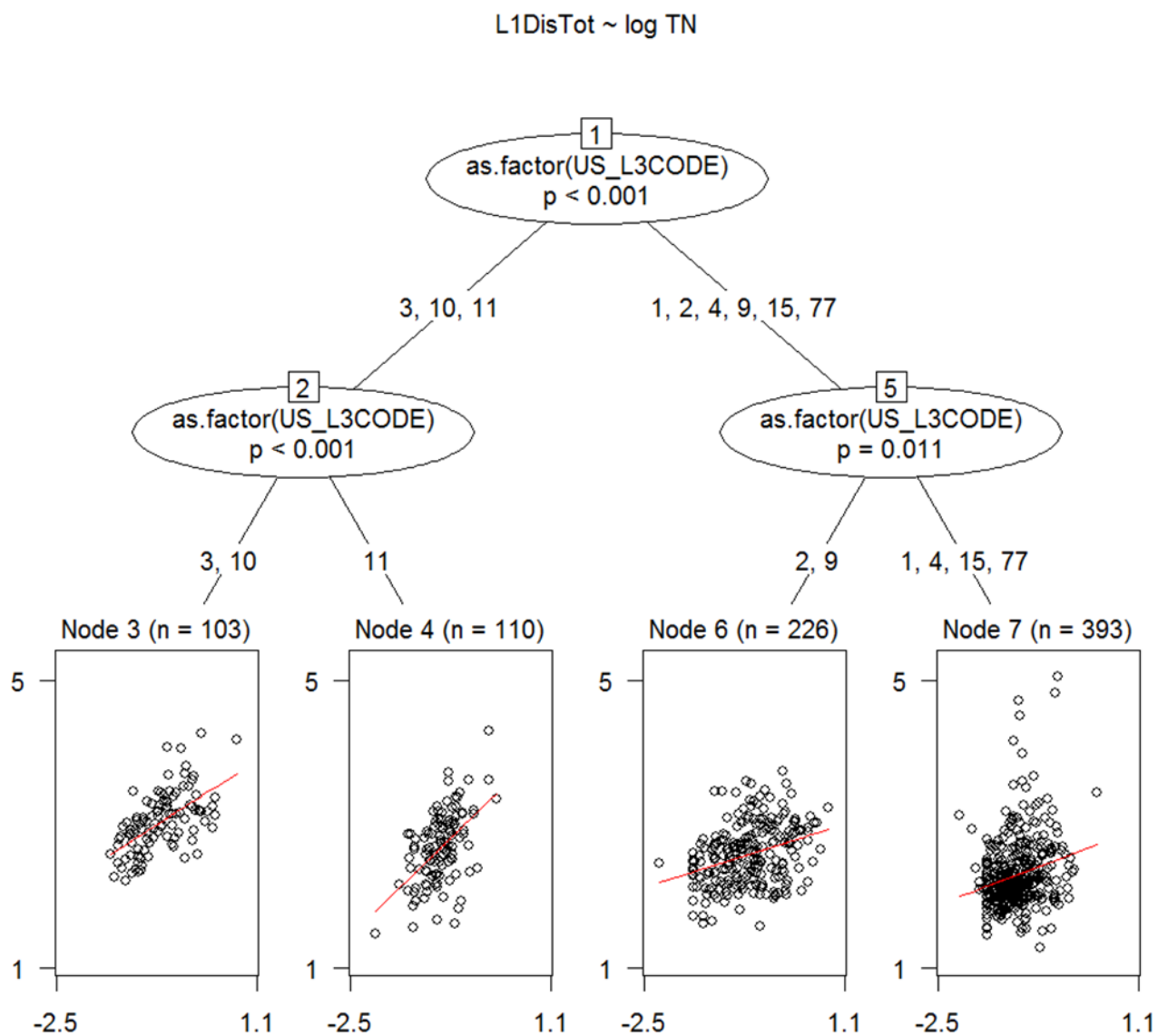


Figure 29 - Model-based recursive partitioning of the Western EMAP diatom optima metric to watershed disturbance (OptCat_L1DisTot) as a response to TN concentration using Omernik Level III ecoregion as the splitting variable.

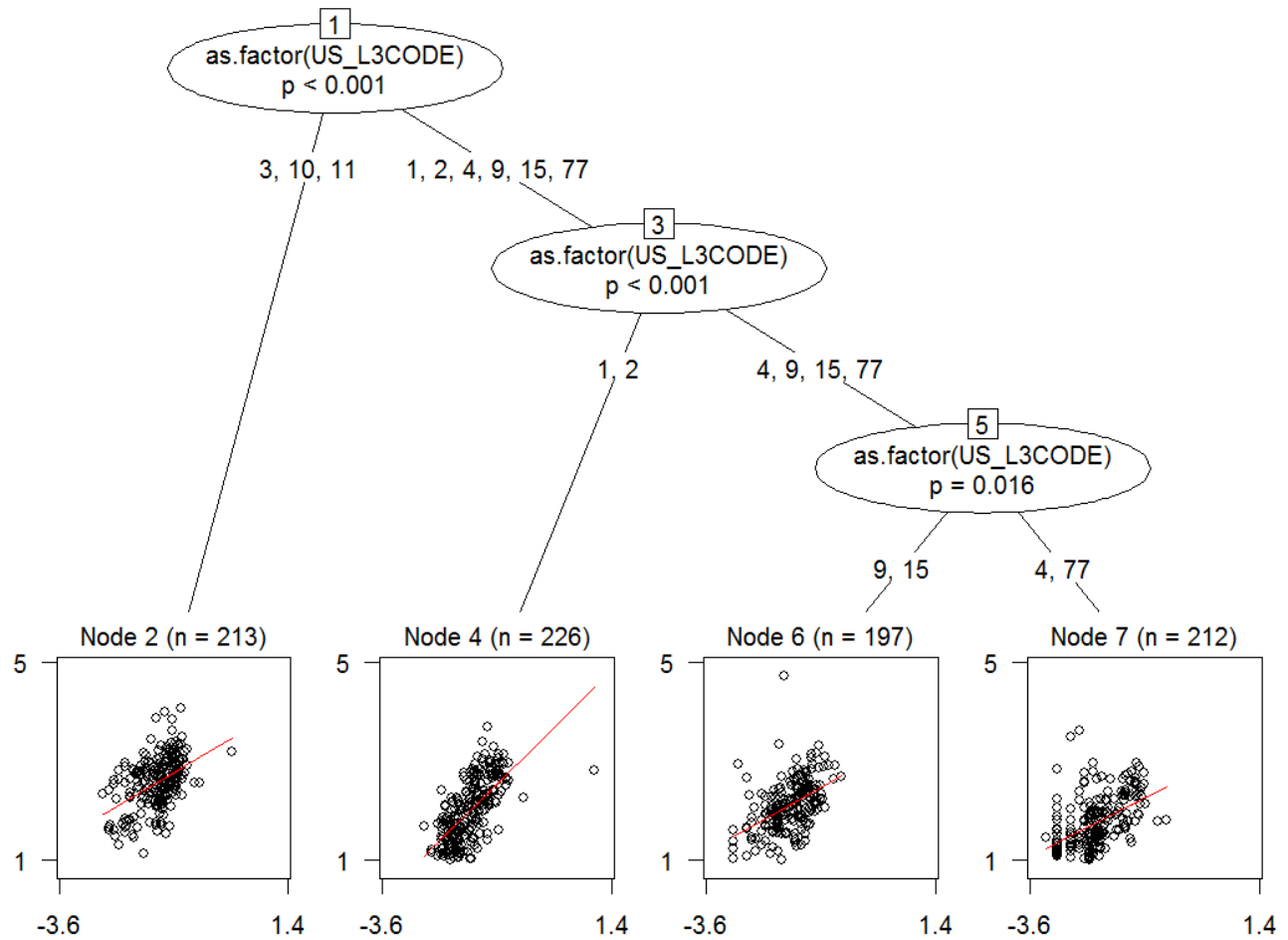


Figure 30 - Model-based recursive partitioning of the Western EMAP diatom MMI periphyton metric (OptCat_NutMMI) as a response to TP concentration using Omernik Level III ecoregion as the splitting variable.

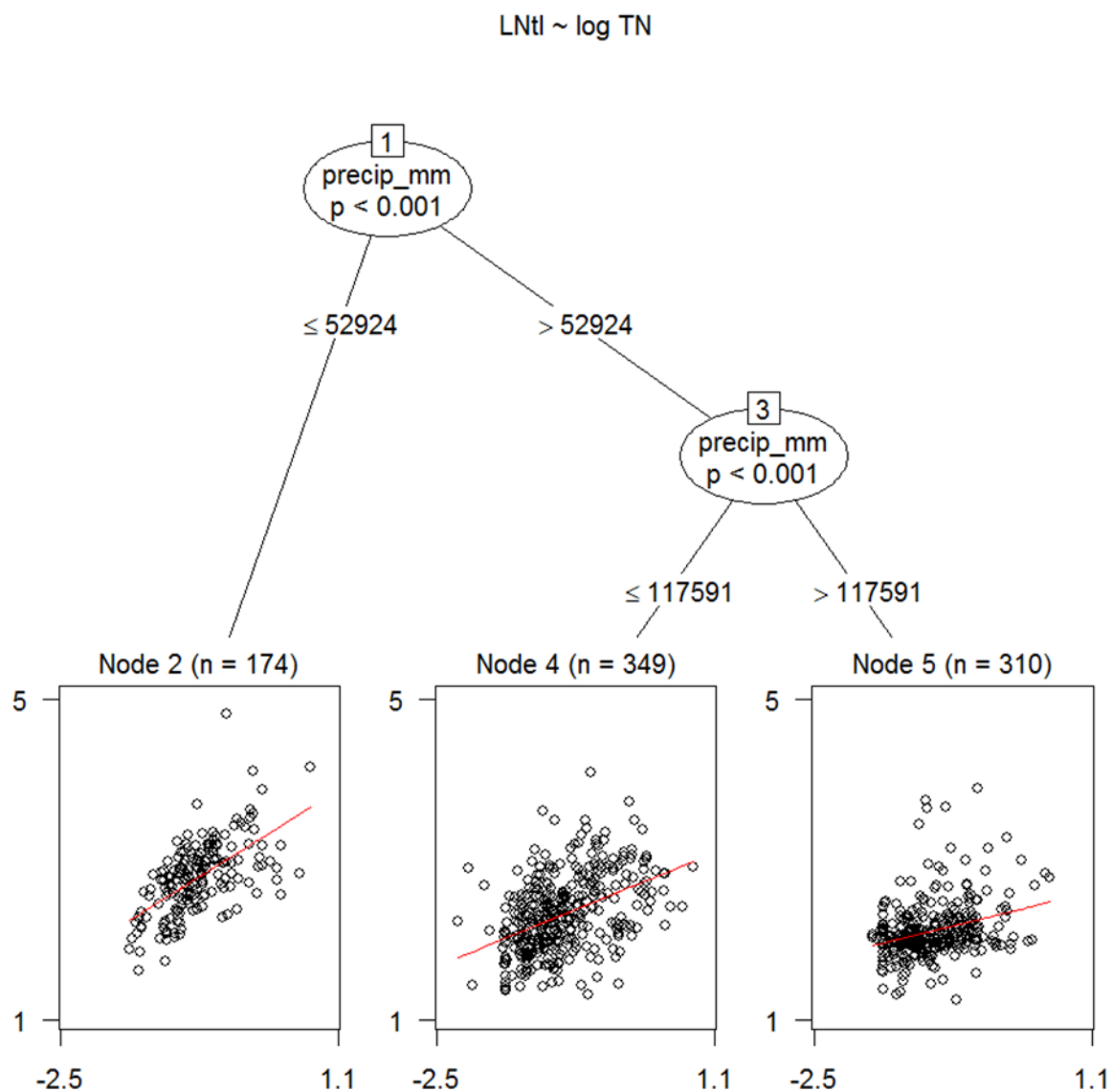


Figure 31 - Model-based recursive partitioning of the Western EMAP diatom optima metric to log TN (OptCat_LNtl) as a response to TN concentration using precipitation as the splitting variable.

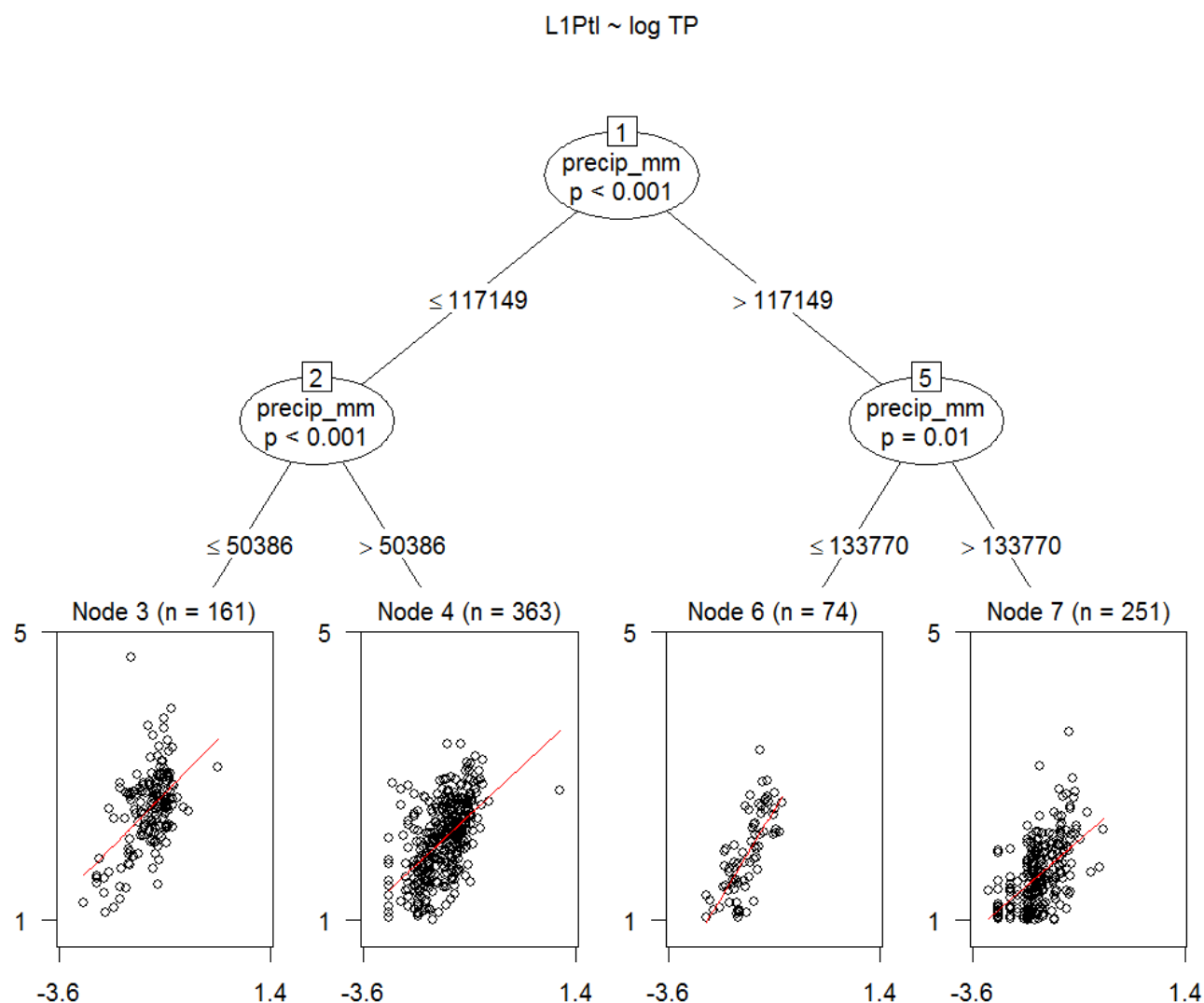


Figure 32 - Model-based recursive partitioning of the Western EMAP diatom optima metric to log TP (OptCat_L1Ptl) as a response to TP concentration using precipitation (units are hundredths of mm) as the splitting variable.

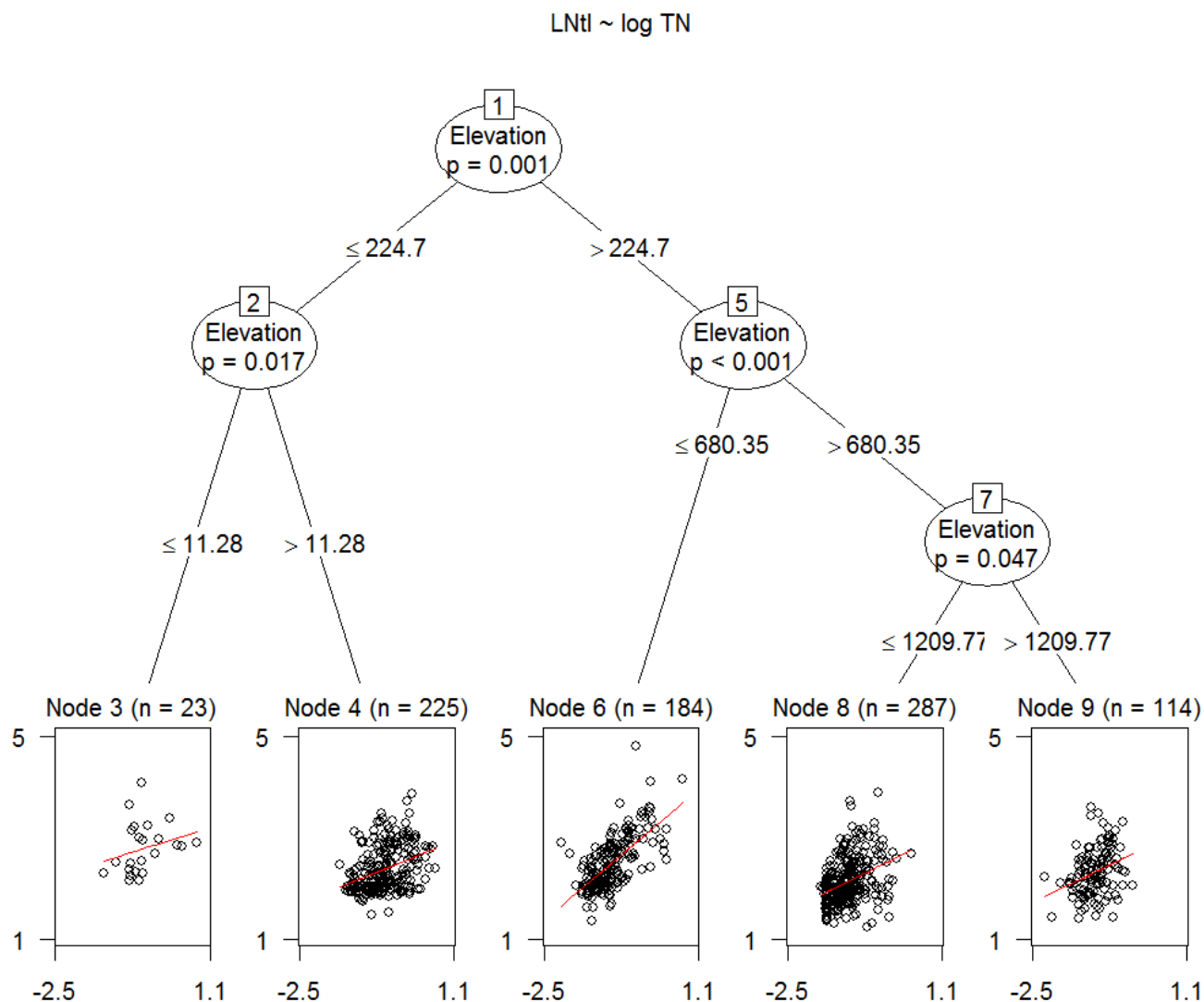


Figure 33 - Model-based recursive partitioning of the Western EMAP diatom optima metric to log TN (OptCat_LNtl) as a response to TN concentration using elevation as the splitting variable.